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Title: Intact orange quality prediction with two portable NIR spectrometers

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Abstract: Two commercial portable spectrometers were compared for orange quality non-destructive predictions by developing Partial Least Squares calibration models, reflectance mode spectra acquisition being used in both. One of them was a Vis/NIR spectrometer in which the radiation reflected by the fruit is collected and conducted by optic fiber to the three detectors (350-2500 nm) of the instrument. The other is an AOTF-NIR with a reflectance post dispersive optical configuration and InGaAs (1100-2300 nm) detector. Four orange varieties were included in calibrations. The parameters studied were soluble solids content, acidity, titratable acidity, maturity index, flesh firmness, juice volume, fruit weight, rind weight, juice volume to fruit weight ratio, fruit colour index and juice colour index. The results indicate a good performance of the predictive models, particularly for the direct NIR prediction of soluble solids content, and maturity index, the prediction of this last parameter being notable for its relevance and novelty. The RPD ratios for these parameters were in the range from 1.67 to 2.21 with the Labspec, which showed better predictive performance, and from 1.03 to 2.33 with the Luminar.

Cover letter

Ttitle: Intact orange quality prediction with two portable NIR spectrometers

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Explanation of the manuscript's significance

In this work, two commercial portable spectrometers of Vis/NIR and NIR were compared for orange non-destructive quality prediction by developing Partial Least Squares calibration models. Four orange varieties were included in calibrations. The parameters studied were soluble solids content, acidity, titratable acidity, maturity index defined by the soluble solids content to titratable acidity ratio, flesh firmness, juice volume, fruit weight, rind weight, juice volume to fruit weight ratio, fruit colour index and juice colour index. The results indicate a good performance of the predictive models, particularly for the direct NIR prediction of the maturity index, soluble solids content, fruit weight and rind weight. The most noteworthy aspects of work are direct NIR predicting maturity index, what is reported as suitable methodology to avoid the difficulty of accurately NIR-predicting titratable acidity.

1 Intact orange quality prediction with two portable NIR spectrometers

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9 Abstract

10 Two commercial portable spectrometers were compared for orange quality non-11 destructive predictions by developing Partial Least Squares calibration models, 12 reflectance mode spectra acquisition being used in both. of One of them was a Vis/NIR 13 spectrometer in which the radiation reflected by the fruit is collected and conducted by 14 optic fiber to the three detectors (350-2500 nm) of the instrument. The other isand an AOTF-NIR with a reflectance post dispersive optical configuration and InGaAs (1100-15 16 2300 nm) detector. were compared for orange non destructive quality prediction by 17 developing Partial Least Squares calibration models. Four orange varieties were 18 included in calibrations. The parameters studied were soluble solids content, acidity, titratable acidity, soluble solids content to titratable acidity ratiomaturity index, flesh 19 20 firmness, juice volume, fruit weight, rind weight, juice volume to fruit weight ratio, fruit colour index and juice colour index. The results indicate a good performance of the 21 22 predictive models, particularly for the direct NIR prediction of fruit weight, rind weight, 23 juice volume, soluble solids content, acidity and the ratio soluble solids content to 24 titratable acidity directly predicted maturity index, the prediction of this last parameter being notable for its relevance and novelty. The RPD ratios for these parameters were in 25

the range from 1.64<u>7</u> to 4.76<u>2.21 with the Labspec, which showed better predictive</u>
performance, and from 1.03 to 2.33 with the Luminar.

Key words: acidity, fruit weight, <u>hardnessfirmness</u>, juiciness, <u>maturity index</u>, NIR,
oranges, soluble solid content.

30

31 INTRODUCTION

Citrus fruit is fast becoming a stable food product in the daily diet of many people. The
 genus Citrus includes several important fruits such as oranges, mandarins, limes,
 lemons and grapefruits, orange and mandarins being the most consumed species. Spain
 leads the world in fresh orange and mandarin exports (Ladaniya, 2008).

Consumers purchase citrus fruits on the basis of quality, it beingas a combination of
 characteristics and attributes are significant for acceptability.

38 Citrus are non-climateric fruit, hence the ripening process stops once separated from the 39 tree and, consequently, fruits can only be harvested and marketed once adequate 40 maturity has been reached (Watkins, 2008). Moreover, the content of sugars and acids 41 in citrus fruit is fairly stable before and after harvest, sugars-to-acid balance being the 42 key to acceptability in these commodities. The content of sugars is generally measured 43 by refractometry as soluble solids content (SSC), sugars representing the main part in it, 44 and acids content is commonly measured as titratable acidity (TA). The ratio of soluble 45 solids content to titratable acidity (SSC/TA) is widely used as maturity criterion for 46 non-climateric fruits (Fellars, 1991), for the reasons indicated above, and particularly 47 used as maturity index in citrus.

48 Organic acids in citrus fruit rank in the 10% range in their contribution to the SSC.

49 Total acidity prediction by Near Infrared Spectroscopy (NIR) has been considered

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50	difficult to achieve, due to the relatively low levels of organic acids in fruits (McGlone
51	et al., 2003; Guthrie et al., 2005; McGlone et al., 2003). Several authors have reported
52	various levels of success in predicting titratable acidity (TA) of pineapple (Shiina et al.,
53	1993), plum (Onda et al., 1994), apple (Sohn et al., 2000), mango (Schmilovitch et al.,
54	2000), intact Imperial mandarin (Guthrie et al., 2005) and, Satsuma mandarin
55	(Hernández et al., 2006), pineapple (Shiina et al., 1993), plum (Onda et al., 1994), apple
56	(Sohn et al., 2000) and mango (Schmilovitch et al., 2000)
57	Citrus fruit are anisotropic objects, showing different physical and chemical properties
58	when measured in different directions. Hence, equatorial measurement are reasonable
59	since at least in citrus, SSC are greatest in the distal apex of fruit decreasing towards the
60	proximal, the opposite happening with A and TA as reported by Peiris et al. (1999).

61

62 Colour is considered as one of the most important external factors of fruit quality 63 (Francis, 1995), since the fruit's appearance greatly influences the consumer. The 64 change of colour in citrus is a consequence of the maturation process, although it is also 65 highly dependent on the existence of cool temperatures at night, not always present 66 under tropical and subtropical growth conditions, which is the reason why a green citrus 67 fruit may or may not be physiologically mature (Olmo et al., 2000).

Fruit softening is often used as criterion for selecting the most suitable harvest date for several commodities (Taylor et al., 1995; Lehman-Salada, 1996). The most common method to determine the firmness of a fruit is destructive and measures its resistance to

71 penetration (Lehman-Salada, 1996; Ahumada and Cantwell, 1996; Mercado-Silva et al., 72 1998). Other methods based on fruit resistance to compression do not necessarily 73 destroy the fruit, but they do require it to beits harvested (Polderdijk et al., 1993; 74 Brovelli et al., 1998). Other methods, in addition to being non-destructive, can be used 75 directly on the tree, such us the use of the hand densitometer (García et al., 1998) or 76 those based on the transmission of acoustic waves through the fruit (Muramatsu et al., 77 1996). In citrus fruit, the relationship between the degradation of the cellular wall and 78 the loss of firmness that accompanies fruit maturation has also been observed (Goto and 79 Araki, 1983). Other important attributes of the internal quality of fruits, along with 80 those mentioned above, are texture and rind thickness. In the same way, juiciness is 81 another important fruit attribute, which can eventually be reduced in citrus by factors 82 affecting the content of the juice sacs, such as freezing or excess of nitrogen fertilization 83 during summer and early autumn (Flint, 1991).

Excepting fruit colour, all <u>of</u> them are impossible to be known by the fruit appearance
by the consumer, whose decision to choose fruit of a desired quality is not supported by
sufficient objective information (Poole et al., 2006).

There is a need for techniques for a swift, non-destructive determination of fruit internal quality, to ensure that all fruit meet a minimum level of acceptance. <u>A-sS</u>implification of the analysis is an important reason for this objective. A further reason is that the conventional destructive methods of analysis are based upon a limited number of fruit samples, whereas non destructive techniques makes it possibleand the possibility to monitor practically all the fruit in real time_are important reasons for this objective. Conversely, improving the environmental sustainability of human activities is a current

94 challenge tothat should be emphasized, achievement of which can contribute to the 95 **<u>nN</u>**on-destructive analytical techniques <u>can contribute</u>, since it does not require 96 chemical reagents or solvents and no waste is generated. 97 The most suitable technology depends on what is the main quality parameter required to 98 be measured. Among several techniques, NIR has great potential for non-destructive 99 determination of internal and maturity attributes (Abbot, 1999). 100 The measurement modes most frequent for intact fruit SSC and TA prediction are-101 reflectance, transmittance and interactance. Although it has been reported predictive 102 outcomes slightly higher using transmittance regarding reflectance and interactance 103 with intact mandarin (McGlone et al., 2003), good results using reflectance mode have 104 been reported with mandarin (Guthrie et al., 2005; Hernández et al., 2006) and orange 105 (Cayuela, 2008). Reflectance is the easiest mode to obtain measurements, since no 106 contact with the fruit is required and light levels are relatively high (Mowat and Poole, 107 1997). In the transmission mode, the measurements are expected to be more influenced 108 by fruit size, the amount of light penetrating the fruit often being very small, thus 109 making it difficult to obtain accurate transmission measurements at grading line speeds 110 (Kawano et al., 1993). 111 NIR technology has been used to determine the soluble solids content (SSC) non-112 destructively in fruit such as apples (Iyo and Kawano, 2001; Hernández et al., 2003; 113 2006), citrus (Tsuchikawa et al., 2003; Guthrie et al., 2005), peaches 114 (Slaughter, 1992; Peiris et al., 1997), cherries (Lu, 2001) and melons (Dull et al., 1989; 115 Dull et al. 1992; Ito et al., 2002; Guthrie et al., 2006), among others. Nicolaï et al. 116 (2007) offered a review on non destructive measurement of fruit and vegetable quality 117 by NIR.

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118 The NIR spectra are the result of the interaction of radiation with the sample, and their 119 physical and chemical properties are reflected in it. Fruit juiciness and fruit weight are 120 fruit physical properties. Successful results of NIR calibrations for citrus juiciness 121 prediction have been reported (Guthrie et al., 2005). The possibility of estimating fruit 122 weight by NIR has rarely been reported, the exploration of this possibility being of great 123 interest, since fruit weight could be added to other fruit quality parameters such as CSS, 124 TA or fruit firmness as different outputs from a single NIR measure. In fact, some good 125 outcomes have been found recently regarding NIR measuring nectarine fruit weight 126 (Pérez et al. 2009). In fact, good results have recently been found in nectarine fruit 127 weight NIR measuring (Pérez et al. 2009). Previous research (Cayuela, unpublished 128 data) also indicated that nectarine fruit weight can be predicted by NIR.

One of the advantages of NIR spectrometry is its portability when the parameter must be measured *in situ*. A few models of portable NIR spectrometers of several brands are available, but the applications to fruit monitoring are few. Furthermore, the technical and constructive characteristics of NIR spectrometers are very diverse, and research into their suitability for use in new applications is needed. Riquelme (2008) included in her doctoral thesis a full revision of the commercial models of on-line NIR instruments and the portable NIR spectrometers applicable to fruit.

In this work, the feasibility of non-destructive NIR prediction of quality parameters on
orange fruit, comparing with two commercially representative portable spectrometers;
using predictive models constructed by Partial Least Squares (PLS), has been evaluated.
The successful prediction of some parameters in this work analysed is reported for the
first time.

141 MATERIALS AND METHODS

142 Fruit

143 Sanguinelli, Valencia, Salustiana and Navelate oranges were hand picked, at random, 144 during the commercial harvest period, from a local experimental grove belonging to the 145 College of Agricultural Engineering, University of Huelva, transported to the laboratory 146 and used immediately or after storage at 4 °C for up to one week. The orange varieties 147 included in this study are taxonomically all them Citrus sinensis (L) Osb. The number 148 of samples from each variety contributing to the set of calibration is indicated in Table 149 1. The oranges were harvested at five different dates from January to April 2009, and 150 therefore, an ample diversity in the fruit quality parameters wereas assured. Before 151 testing, samples were taken out from the cooling and maintained at room temperature 152 (23-25 °C) for 18 h in order to allow for acclimatization to the experimental conditions. 153 Fruit were cleaned with a cloth moistened in sterile water, then dried at the lab 154 environment prior to measurement. Each fruit unit constituted the sample, and were 155 numbered in the fruit peduncle area.

156 Spectral acquisition

The spectral acquisition of every sample was performed using two portable
spectrophotometers with different optical and constructive features: Labspec (Analytical
Spectral Devices Inc., Boulder) and Luminar 5030 (Brimrose Corp., Maryland).

160	Labspec is a Visible/NIR spectrometer equipped with three detectors. The detector for	r
161	the visible range (350-1000 nm) is a fixed reflective holographic diode array with	a

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162	sensitivity of 512 pixels. Wavelengths in the visible spectrum can carry information
163	relating to some of the quality parameters analyzed, such as fruit color and color of the
164	juice, and perhaps additional ones as the fruit size. The portability of the equipment is
165	assured by the weight of the spectrometer 8.5 kg. The wavelength range of 1000-1800
166	nm is covered by a holographic fast scanner InGaAs detector cooled at -25°C. The same
167	aforementioned device coupled with a high order blocking filter operates for the 1800-
168	2500 nm interval. The instrument is equipped with internal shutters and automatic offset
169	correction, the scanning speed being 100 ms. The acquisition of spectra was performed
170	using the high intensity contact probe accessory of the spectrometer, with light source
171	diameter 20 mm-(Fig. 1), and standard SMA 905 fiber optic connectors. The whole
172	spectrum (350-2500 nm) was acquired, each spectral variable corresponding to 2 nm
173	interval. The repeatability of the instrument, expressed as standard deviation on the
174	average absorbance of 350 to 2500 nm of five measures of a white tile, is 6.00 10 ⁻⁴ . The
175	orange spectra acquisition was carried out using Indico Pro software (Analytical
176	Spectral Devices Inc., Boulder). The portability of the equipment is assured by the
177	weight of the spectrometer 8.5 kg.

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Luminar 5030 is an AOTF (acousto-optic tunable filter) NIR spectrophotometer, equipped with a reflectance post dispersive optical configuration and InGaAs (1100-2300 nm) detector. The reference spectrum is automatically taken by the instrument, just as an UV-VIS spectrophotometer dual beam; the beam is divided before leaving and a small portion is sent to a second detector that makes the reference. The scanning speed in Luminar 5030 is 60 ms. The spectrometer is formed by the hand-held unit, shaped and used similarly to a 'gun', the diameter of the exit cone of the light source

185	being 8 mm (Fig. 1), and the computer unit; the spectrometer set, with a total weight of
186	5.26 Kg, offers a good portability, with 4h of autonomy using a set of batteries that
187	allows in situ measuring on crop; this is an important advantage to note. The hand-held
188	unit is equipped with a base for facultative use in the laboratory. The whole spectrum
189	(1100-2300 nm) was acquired, each spectral variable corresponding to a 2 nm interval
190	The repeatability of the instrument, expressed as standard deviation on the average
191	absorbance of 1100 to 2300 nm of five measures of a white tile, is 6.76 10-4. The
192	signals were acquired with software Acquire (Brimrose Corp., Maryland).

193 Citrus fruit are anisotropic objects, showing different physical and chemical properties 194 when measured in different directions. Equatorial measurement was selected because, at 195 least in citrus, SSC are greatest in the distal apex of fruit, decreasing towards the 196 proximal, the opposite happening with A and TA, as reported by Peiris et al. (1999). 197 198 The sample unit was the fruit, an averaged spectrum being obtained for each sample, 199 resulting from a total 100 spectra corresponding to two measures of 50 spectra each, it 200 for both spectrometers, taken at opposite equatorial locations, according to the reasons 201 indicated above.

202

Figure 1

203 Reference analysis

The quality attributes of each fruit were evaluated by analysis of their physical and
 chemical parameters. Additionally, parameters calculated arithmetically were assessed.
 <u>Physical parameters</u>

207 Fruit weight (FW, g) and rind weight (RW, g) were measured using electronic precision 208 balance (0.001 g). RW was determined once the fruit was pealed by separating the rind 209 from the flesh. For the measurement of intact fruit colour a spectral colour analyser 210 (colorimeter PCE-RGB 1002) was used, determining the same in triplicate for the same 211 fruit. This instrument has a RGB (red, green, blue) colour scale 0-1023. The Easy-RGB 212 software (Logicol Colour Technology Co.) was used for the conversion into the L, a and 213 b parameters of the Hunter scale. The results expressed by the colour index (FCI) were 214 obtained from the mathematical formula [1].

[1]

215 CI= 1000a /Lb

216 Flesh firmness (F, KgN) was quantified using a hand penetrometer (TR FT-327 Turoni 217 S.r.l., Forcy, Italy) with a 7 mm diameter cylindrical plunger, twice on the peeled fruit 218 at the equatorial circumference. The fruit were halved through the equatorial plane and 219 juice extracted with a commercial juice extractor. Juice volume (JV, mL) was measured 220 with a test tube. Juice volume to fruit weight ratio was calculated and expressed as a 221 percentage (JV/FW, %). Juice colour was determined with the same colorimeter and 222 units above indicated for fruit colour and expressed as juice colour index (JCI). The 223 measurement was performed on a juice sample of each individual orange fruit in a Petri 224 plate, this measurement being made through the glass at the bottom of the plate. This 225 procedure was used to avoid the risk of any introduction in the juice of the colorimeter 226 light source. The SSC was measured on each fruit juice by a hand help digital 227 refractometer (Atago Co, PAC 1 Brix Meter, Tokyo) and obtained from two replicates, 228 expressed as Brix units. 229

230 Chemical parameters

232	Co, PAC-1 Brix-Meter, Tokyo) and obtained from two replicates, expressed as
233	percentage.
234	The acidity (A, pH) was measured on the juice of each individual fruit from two
235	replicates, using a digital pH-meter. The titratable acidity (TA) was analogously
236	measured from two replicates by direct titration of a 10 mL juice sample added with 10
237	mL distilled water, neutralized with NaOH 0.1 N until pH 8.2 and expressed as citric
238	acid (g/L).
239	Arithmetically calculated parameters
240	The maturity index reference (MI) ratio soluble solids content to titratable acidity-was
241	arithmetically obtained from the ratio between SSC and TA reference analysis values
242	(SSC _R /TAR) and in turn, by <u>t</u> he arithmetical computation from SSC and TA obtained
243	using the NIR predictive models developed in this work for both parameters
244	(SSC_P/TA_P) was compared with the prediction outcomes from the model developed for
245	MI directly predicting-
246	
247	Chemometrics and calibration procedure

231 The SSC was measured on each fruit juice by a hand-help digital refractometer (Atago

Partial Least Squares (Wold et al., 1983) models were obtained with Unscrumbler 9.7
(CAMO Software AS, Norway). For the Labspec spectra, noise intervals 350-499 nm
and 2301-2500 nm were removed. Calibration tests were conducted for the Labspec to
eliminate spectral noise at the beginning and end of spectrum, using 500 to 2300 nm
wavelengths for this purpose. In turn, tests were carried out for the same spectrometer
excluding 600-750 nm, a range strongly affected by the skin pigment chlorophyll that
absorbs red light, whose absorbance band corresponds to 680 nm-(McGlone et al.,

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255 2003). This band is not included in the Luminar's wavelengths range. Exclusion of the
256 initial and final areas of spectrum to eliminate noise, in the case of Luminar was also
257 considered unnecessary. Faulty reference analytical data were identified and eliminated
258 prior to calibration.

259 Before calibrations, the reflectance data were transformed to absorbance, mean 260 normalized, and optionally treated by multiplicative scatter correction (MSC) using 261 Unscrumbler 9.7. The influence of the pre-processing on the prediction of the 262 calibration models was tested by different gap and smooth combinations for first and 263 second gap-segment derivatives. Standard <u>Nn</u>ormal <u>V</u>ariate <u>t</u>ransformation (SNV) 264 was also tested. Full-cross internal validation (FCV) was used for building the models. 265 Calibration tests were also conducted with different numbers of principal components in 266 order to determine the number of PCs optimum and the results assessed in terms of 267 standard error of cross validation (SECV). Exceptionally, where indicated in the tables, 268 points clearly separated from the calibration sets in the scatter plots were identified as 269 outliers and removed with Unscrumbler 9.7 using its specific application for this 270 purpose.

271

272 Validation procedure

Two Eexternal validation exercises were carried out using the corresponding models for
predicting the parameters on completely independent samples, One validation exercise
was conducted using 1/5 being reserved from the total reference analysed of the total
number of samples for each parameter, the set for validation being formed by
Unscrumbler's specific application, the first from every 5 samples taken with this
purpose (V1). Other exercise was conducted for the set of samples 51 to 100,

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279 <u>corresponding to Sanguinelli (V2), which was excluded from the corresponding</u> 280 <u>calibrations that were developed previously for this purpose.</u>

In assessing the soundness of the calibrations performance, the main considerations were the root mean square error of prediction (RMSEP) and the residual predictive deviation (RPD), described by Williams and Sobering (1996) as the ratio of the standard deviation of the reference data for the validation set to the SEP.

285 Paired samples T test for dependent samples were also conducted to verify the results 286 from RPD and RMSEP analysis. Not established as statistic specifically for assessing 287 PLS model performance, T test was applied here exclusively referred to the data pairs 288 included in the external validation exercise. For this purpose, data pairs of the reference 289 value and the resulting prediction corresponding to the external validation sets were 290 compared. The paired T test is a parametric procedure, useful for testing whether the 291 means of two groups are different, where the samples are drawn in pairs. The T test was 292 carried out using SPSS Statistics software (SPSS Inc., Chicago). The compliance with 293 the null hypothesis in this test (*P>0.05) indicates that the measure NIR provides at 294 least the same accuracy as the reference method.

The prediction output from the calibration model for direct NIR <u>MI</u> measuring the SSC/TA ratio was compared both with SSC_R/TA_R the reference values and with SSC_P/TA_P <u>above</u> described <u>above</u>, also using the paired samples T test.

298

299 **RESULTS**

300 Fruit NIR spectrum

301	Typical diffuse absorbance spectra for intact oranges acquired by both instruments
302	(Labspec and Luminar 5030) are shown in Fig 21. The spectra from the two instruments

showed characteristics similar to those also previously described in oranges (Cayuela,
2008) or in Satsuma mandarin (Hernández et al., 2006) beyond differences regarding
the wavelengths acquired.

306 High intensity peaks are noted, related to the strong water absorbance bands present 307 from their first overtone at 1400 to 1500 nm and combination band at 1880-2100 nm 308 with an interval of relative intensities of high absorption, approximately between 1400 309 and 2200 nm. Absorbance falls around the 1500 nm and rises again from approximately 310 1850 nm where oscillation exists probably due to carbohydrates. Overall, the pattern of 311 the absorption curves is similar to that for other fruit such as peach (Peiris, 1997), 312 kiwifruit (McGlone and Kawano, 1998), mangoes (Budiastra et al., 1998), apple (Lu et 313 al., 2000) and cherries (Lu, 2001).

314315

Figure 21

316 **Population characterization**

317 The mean and standard deviation values of the populations of orange varieties used in 318 the calibrations and the external validations for the parameters analysed are shown in 319 Table 1. The ratio SSC/TA-MI in the different varieties shows that Valencia oranges 320 corresponded to a less advanced stage of maturation, Sanguinelli and Navelate being at 321 approximately the same maturity stage and Salustiana showing, according to its 322 SSC/TA-MI ratio, the most advanced maturity of the four varieties. As can be seen, 323 wide ranges of variation of all the parameters analysed were included in the 324 populations.

The characteristics of the sample sets used for the validation exercises, conducted using obtained models for predicting each parameter in samples independent from the calibrations sets, are shown in Table 2. The contribution of each variety to the 328 validation sets were 50 Sanguinelli, 10 Valencia, 9 Salustiana and 10 Navelate samples,

Table 2

329 according to the proportion 1/5 regarding calibration sets.

330 Table 1

331

332 Calibration development

The statistical coefficients of best calibration models with Labspec and Luminar 5030for predicting orange quality parameters according to the treatment used for the two

instruments tested, are included in Tables 3 and 4 respectively.

The range from 600 to750 nm could be affected by the skin pigment chlorophyll, as has been previously indicated. Calibration tests were conducted with the Labspec excluding this range, without any improvement when including it. This result probably reflects the fact that no green areas were present in the skin of any of the orange samples used in the calibrations. This test was unnecessary with the Luminar, since its spectral range does not include 600-750 nm.

342 The calibration coefficients statistics for V1 were relatively similarclose with both 343 instruments for SSC, A, SSC/TAMI, JV and JV/FW. Hence, with the Labspec the 344 RMSECV were 0.60, 0.12, 1.81, 7.39 and 0.03, whereas for the Luminar it were 0.67, 345 0.15, 2.06, 7.97 and 0.03 respectively for these parameters (Table 3 and Table 4). Some 346 differences were found between both spectrometers for FW, RW and FF, the Labspec 347 showing RMSECV 19.89, 14.61 and 1.18, whereas for the Luminar it were 22.69, 17.07 348 and 1.27 (Table 3 and Table 4). All the cases, the lower values RMSECV from the 349 Labspec implies a better predictive performance for these parameters. Calibration for 350 FCI and JCI were attempted only with the Labspec, since it integrates Visible and NIR.

351	Generally, mean normalized data provided the best fits for most parameters analysed. In		
352	some cases, shown in Tables 3 and 4, in both or either spectrometers, MSC treatment		
353	alone or after mean normalization facilitated the best calibration coefficients.		
354	Table 3		
355	Table 4		
356	Model external validation		
357	The statistical coefficients of the predictions carried out inoutcomes from the external		
358	validation exercises <u>V1</u> for each orange quality parameter with both spectrometersare		
359	shown in Table 3 and Table 4. The predictions output versus the analysed value of each		
360	parameter in th <u>ise</u> external validation exercise are shown in Fig <u>32</u> .		
361	As can be seen from the validation plots, and according to the validation coefficients		
362	RMSEP and RPD shown in Table 3 and Table 4, prediction accuracy was similar for		
363	most fruit quality parameters with both spectrometers and slightly better with the		
364	Labspec for FW, RW and FF.		
365	The validation exercise V2, corresponding to Sanguinelli, provided better performance		
366	for all the parameters analyzed with the Labspec. The statistical values are shown in		
367	Table 5. The best RPD with the same spectrometer in this exercise were reached for		
368	SSC and MI predictions.		
369	Figure 3 2		
370	<u>Table 5</u>	<	Formatted: Font: (Default) Arial, Not Bold
371	DISCUSSION		Formatted: Centered
372	The predictive calibration for SSC showed slightly better fit with the Labspec, although		
373	it was very close to that obtained with the Luminar 5030, in which the external		

validation exercise V1 carried out for SSC showed analogous accuracy with both	
spectrometers, as can be deduced from the RPD ratios. gave a more accurate prediction	
- Acidity, Titratable Acidity and the SSC/TA ratio predictive calibrations were also very	
close between both spectrometers.	
The values RMSEP in the predictive models-validations V1 and V2 for SSC 0.74 and	
0.87 (Labspec) or 0.68 and 1.12 (Luminar) were slightly higher to those previously	
reported by Cayuela (2008) also for Valencia Late orange and using reflectance (570-	
1850 nm) with a <u>non portable Vis-NIR spectrometer InfraXact</u> (Foss), where RMSEP	
0.55 is reported. In that work the direc NIR prediction of MI was not addressed.	
For A, TA and MI the accuracy were similar in the V1 validation exercise for the	
Labspec (Table 3) and the Luminar (Table 4). The validation V2 conducted with a	
Sanguinelli fruit set (Table 5) showed lower RPD values for all these parameters and	
both spectrometers, excepting for MI with the Labspec. The RPD values were higher for	
the Labspec in this exercise, hence showing better performance.	
Both predictive calibrations for acidity (pH) and titratable acidity in this study with both	
spectrometers achieve lower RMSEP values than the 0.33 for titratable acidity and 0.49	
for pH reported previously with oranges (Cayuela, 2008), which could be attributed to	
the wider ranges of the values of these parameters included in the calibrations in this	
work. For a Fantec-FQA, from the SEP and σ for pH and TA reported by Riquelme	Formatted: Font: Symb
(2008) can be deducted RPD values 0.68 and 0.56 using internal software for mandarins	
and oranges, as well 0.79 and 0.37 for the same fruits using calibration developed by the	
author, these values being lower to those reported in this work.	
Several other results have been reported with Satsuma mandarin (Hernández et al.,	
	validation exercise_ <u>V1</u> <u>carried out for SSC showed analogous accuracy with both</u> spectrometers, as can be deduced from the RPD ratios, gave a more accuracy with both -Acidity, Titratable Acidity and the SSC/TA ratio predictive calibrations were also very elose between both spectrometers. The values RMSEP in the predictive models validations V1 and V2 for SSC 0.74 and 0.87 (Labspec) or 0.68 and 1.12 (Luminar) were slightly higher to those previously reported by Cayuela (2008) also for <u>Valencia Late</u> orange and using reflectance (570– 1850 nm) with a <u>non portable</u> Vis-NIR spectrometer InfraXact (Foss) _a where RMSEP 0.55 is reported. <u>In that work the direc NIR prediction of MI was not addressed</u> . For A, TA and MI the accuracy were similar in the V1 validation exercise for the Labspec (Table 3) and the Luminar (Table 4). The validation V2 conducted with a Sanguinelli fruit set (Table 5) showed lower RPD values for all these parameters and both spectrometers, excepting for MI with the Labspec. The RPD values were higher for the Labspec in this exercise, hence showing better performance. Both predictive calibrations for acidity (pHI) and titratable acidity in this study with both spectrometers achieve lower RMSEP values than the 0.33 for titratable acidity and 0.49 for pH reported previously with oranges (Cayuela, 2008), which could be attributed to the wider ranges of the values of these parameters included in the calibrations in this work. For a Fantec-FQA, from the SEP and σ for pH and TA reported by Riquelme (2008) can be deducted RPD values 0.68 and 0.56 using internal software for mandarins and oranges, as well 0.79 and 0.37 for the same fruits using calibration developed by the author, these values being lower to those reported in this work. Several other results have been reported with Satsuma mandarin (Hernández et al.,

2006) also using reflectance and with Imperial mandarin using other optical modes 397

ool

398 (Guthrie et al., 2005; McGlone et al., 2003). However, a comparison is not 399 recommended because oranges are larger in size, and withhaving a different structure to 400 mandarin fruit. Moreover, literature about using NIR spectroscopy to SSC and pH 401 measuring in several fruit is very wide, as is referred in the introduction, thus the aim 402 here is not to demonstrate this possibility and their industrial application.

403 The RPD coefficients from the external validations for SSC, A and TA were comprised 404 in both spectrometers between 1.540.80 and 2.33. For reference, RPD ranging from 405 0.12 to 0.56 were reported using a Fantec-FQA for the same parameters (Riquelme, 406 2008). Models constructed with both spectrometers were able to determine SSC and A 407 in V1 with an accuracy approaching 95%, according to an arithmetic calculation of the 408 difference between the values estimated by NIR and reference values (data not shown). 409 The external validation V1 with the calibration for the ratio SSC/TAMI showed RPD 410 very similar values in both spectrometers, this being 1.67 with the Labspec and 1.64 411 with the Luminar. The exercise V2 showed for the same coefficient RPD with the 412 Labspec 1.75, higher to the RPD 1.26 with the Luminar. Riquelme (2008) reported 413 RPD 0.25 for the same parameter and also in oranges using a Fantec-FQA., but not 414 much information scientific has been reported regarding citrus MI direct NIR 415 prediction. 416 The predictions obtained in the exercises V1 and V2 with the models for NIR 417 measuring directly Soluble Solids Content to Titratable Acidity ratio (SSC/TA)MI, were 418 compared using paired sample test T both with the ratio SSC/TA arithmetically 419 obtained from SSC and TA reference analysis values (SSCR/TAR) and with the 420 arithmetical computation from SSC and TA independient NIR obtained (SSC_P/TA_P). No 421 significant differences were found in V1 between SSC/TAMI predictions and

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422	SSC_R/TA_R references (sig. 0.629) while on the contrary, there were significant
423	differences between SSC _P /TA _P and SSC _R /TA _R references (* $P = 0.039$). Similar result was
424	found with V2 exercise (data not shown). Therefore, the ratio SSC/TA arithmetically
425	obtained from SSC and from TA valuesMI is easy and accurately if it is obtained from
426	the reference analysis values (SSC_R/TA_R) , but the arithmetical computation of MI
427	would give worse results if these values are to be NIR obtained (SSC _P /TA _P), which will
428	be the case when using a NIR spectrometer in routine analysis. The <u>mrefore</u> , the use of
429	models for MI direct NIR prediction is advantageous. As the SSC/TA ratio constitutes
430	the most widely used maturity indexMI, the possibility of its NIR prediction in real time
431	is of great interest, regardless of the SSC and TA values, which are also NIR measured
432	at the same moment.
433	On the other hand, th <u>ise</u> study conducted revealed that NIR prediction of <u>SSC/TAMI</u> , a
434	more interesting maturity index than titratable acidity by itself, were more accurate than
435	TA NIR prediction, as revealed in the V1 exercise by the statistical coefficients of the
436	calibration and external validation exercises RPD for both parameters SSC/TAMI and
437	TA, included in Table 3 (Labspec) and Table 4 (Luminar), and in the V2 exercise
438	(Table 5) for both spectrometers. From these results it appears that it is a methodology
439	suitable for avoiding the difficulty of accurately NIR-predicting titratable acidityTA, as
440	described by various authors (Mc Glone et al., 2003; Guthrie at a., 2005), since it is
441	possible to predict SSC/TAMI ratio directly by NIR ₇ . which could be used as maturity
442	index .
443	
444	<u>+</u> <u>T</u> he high accuracy of the predictive calibrations for fruit weight and rind weight <u>in the</u>

445 <u>V1 exercise</u> must be highlighted since, for the first time, it makes it possible to measure

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446	these parameters by NIR spectroscopy in citrus. Fruit weight NIR measuring is very
447	scarcely mentioned in the literature, even for other fruit, as indicated in the introduction.
448	This accuracy is referred to each weight measurement on single fruit, it being essential
449	for the industry to ensure that all the fruit included in each bag or box has a minimum
450	suitable size or weight. The RPD from the external validation reached with the Labspec
451	for fruit weight and rind weight in the V1 exercise were 4.76 and 4.54, and with the
452	Luminar, 3.03 and 3.23, values, all cases being higher than three, the value that ideally
453	should be reached to ensure the goodness of a model (Williams and Sobering, 1996).

455 Regarding Juice volume calibrationsmodels in V1 validation, they were fairly accurate, 456 coefficients being very close in both spectrometers, as shown in Tables 3 and 4. The 457 external validation for juice volume with bB oth spectrometers showed RPD near to 458 three in this exercise, revealing good performance. Although the volume of orange juice 459 is not generally included in trade rules for citrus fruits, the accuracy level needed has 460 not even been defined, nevertheless it is a very interesting quality parameter to be 461 monitored in each single fruit, as Riquelme (2008) reported, since citrus can eventually 462 suffer from juiciness defects.

463

454

The external validation <u>V1 with the calibration</u> for <u>the</u> ratio JV/FW with both <u>the</u> Labspec (Table 3) and <u>the</u> Luminar (Table 4) RPD showed <u>RPD</u> values higher to 1.5 and lower to 2, what are not optimal but reveals a good predictive potential of these models. The statistical coefficients from the external validation exercises corresponding to each of these parameters with both spectrometers are shown in Table 3 and Table 4.

470 Fruit flesh firmness showed the valueRPD 1.85 in the V1 external validation with the 471 Labspec. The ability of NIR estimating orange flesh firmness with sufficient accuracy is 472 also of great interest, since it could be used as maturity index (Olmo, 2000), although 473 more research is needed about the relationship between orange firmness and maturity 474 stage to establish maturity indices based upon this parameter on specific varieties. The 475 accuracy level reached in the external validation V1, with the Labspec for flesh firmness 476 was 83.9%, and 79.0% with the Luminar, according to an arithmetic calculation of the 477 difference between the values NIR estimated and reference values (data not shown). 478 This means that the difference between the model prediction and the reference analysis were, in both cases, approximately in the order of 20%. 479

480

481 The external validation <u>V1</u> for FCI corresponding to the Labspec provided low and 482 relatively good accuracy RMSEP values, with RPD 1.75. This last data indicate 483 differences between model prediction and reference analysis of approximately 10%, on 484 average.

The statistical coefficients of the external validation <u>V1</u> for predicting JCI (RPD 1.31) revealed that model performance was far from good. However, the paired T test determined no significant differences between predictions and reference values. This fact could be related with some possible inconsistent reference values that may be attributed to the methodology used for measuring juice colour and, particularly, the inclusion of the Petri plate bottom in the juice colour measurement, needing to be improved.

492 However, considerable disagreement was found in the statistics between both validation
493 exercises V1 and V2 conducted with both spectrometers for FW, F and JV. The RPD

494	values were clearly lower for these parameters in the exercise V2. The RPD reached for
495	the JV/FW, FCI and JCI in the V2 exercise were also lower than in V1. This facts
496	shows that the predictive potential of the technique for these parameters, when
497	measuring a set of samples not considered in the calibrations, can be lower than
498	expected.
499	
500	Some Most quality orange fruit parameters showed a RPD ratio lower than three, value
501	considered as threshold for an optimal validation (Williams and Sobering, 1996). These
502	parameters were soluble solids content, acidity, titratable acidity, soluble solid content
503	to titratable acidity ratio, flesh firmness, juice volume to fruit weight ratio, and
504	additionally, with the Labspec colour index and juice colour index. However,
505	consideration of other criteria in the external validation exercises, such as paired test T
506	for all parameters studied indicated no significant differences between the laboratory
507	reference and the NIR determined values, which means that NIR was at least as accurate
508	as the reference methodologies. The unique particularity was with colour index,
509	presenting *P 0.055, which is the limit of significance, therefore reference and NIR
510	measure were very near to be different in this parameter.
511	The predictions for fruit weight, rind weight and flesh firmness showed slightly better
512	statistical coefficients from calibration and external validation using the Labspec- This
513	fact can be due to the contact probe used in spectra acquisition with theis
514	Labspeespectrometer, presenting light source diameter (20 mm), larger than that of the
515	Luminar 5030 (8 mm), and, therefore, being different the area of the fruit surface
516	illuminated in both cases. This factor contributes to being able to explore a larger
517	portion of the fruit in the case of the spectrometer Labspec.

CONCLUSIONS

519	The predictions of both SSC and TA and the SSC/TA relationship has been successful
520	in the external validation by applying the corresponding calibrations developed using
521	both spectrometers. The direct NIR prediction of SSC/TA ratio was advantageous
522	regarding the arithmetical computation from the values of SSC and TA NIR obtained.
523	The NIR prediction of SSC/TA was more accurate than TA NIR prediction. Therefore
524	NIR directly predicting SSC/TA ratio can be a suitable methodology to avoid the
525	difficulty of accurately NIR predicting titratable acidity.
526	The predictive calibrations for Fruit Weight and Rind Weight facilitated high accuracy
527	with both spectrometers, which is the first time that measuring these parameters by NIR
528	spectroscopy has been reported in citrus. Also, for Juice volume and JV/FW ratio
529	calibrations were fairly accurate.
530	
531	The use of NIR measurements of Fruit Colour Index together with the ratio SSC/TA
532	and with the Flesh Firmness predictions may contribute to better defining the ripeness
533	of citrus fruit regarding that obtained with these parameters separately.
534	The satisfactory result obtained in the validation of models integrating four orange
535	varieties showed that it is possible to develop predictive models for orange quality
536	parameters in general, and not only models specific for an orange variety.
537	

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544 **REFERENCES**

- Abbot, J.A., Massie, D.R., Upchurch, B.K., Hruschka, W.R., 1995. Non-destructive
 sonic firmness measurement of apples. Transations of the ASAE, 38, 1461-1466.
- 547 Abbot, J.A. 1999. Quality assurement of fruits and vegetables. Postharvest Biol.
 548 Technol. 15, 207-225.
- 549 Ahumada, M., Cantwell, M., 1996. Postharvest studies on pepino dulce (*Solanum muricatum* Ait.): maturity at harvest and storage behaviour. Postharvest Biol. Technol.
 551 7, 129-136.
- 552 Armstrong, P.R., Brusewitz, G.H., Stone, M., Brusewitz, G.H., 1997. Non-destructive
- 553 acoustic and compression measurements of watermelon for internal damage detection₂-
- ⁵⁵⁴ <u>Hin: NRAES (Ed.)</u>, Sensors for Non-destructive Testing International Conference and
 ⁵⁵⁵ Tour, <u>18-21 February 1997</u>, Orlando, Florida, <u>pp. -172-182.</u>
- Blankenship, S.M., Parker, M., Unrath, C.R. 1997. Use of maturity indices for
 predicting post-storage firmness of Fuji apples. HortSci. 32, 909-910.
- 558 Brovelli, E.A., Brecht, J.K., Sherman, W.B., Sims, C.A., 1998. Potential maturity
- 559 indices and developmental aspects of melting-flesh peach genotypes for fresh market. J.
- 560 Amer. Soc. Hort. Sci. 123, 438-444.
- 561 Budiastra, W., Ikeda, Y., Nishizu, T., 1998. Prediction of individual sugars and malic
- acid concentrations of apple and mangoes by NIR reflectance system. Journal of
 JSAM_{7.} 60, 117-12.
- 564 Cayuela, J.A. 2008. Vis/NIR soluble solids prediction in intact oranges (*Citrus sinensis*565 L.) cv. Valencia Late by reflectance. Postharvest Biol. Technol. 47, 75-80.
- 566 DeEll, J.R., 1996. Chlorophyll fluorescence as a rapid indicator of postharvest stress in
- 567 apples. Ph. D. thesis, University of Guelph, Guelph, Ontario, Canada.

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- 568 Dull, G.G. Birth, G.S. Smittle, D.A. Leffler, R.G. 1989. Near Infrared Alaysis of
- 569 soluble solids in iuntate Cantaloupe. J. of Food Sci. 54, 393-395.
- 570 Dull, G.G. Leffler, R.G. Birth, A. Smittle, A. 1992. Instruments for non destructive
- 571 measurement of soluble solids in honeydew melons. Trans. ASAE. 35, 735-737.
- 572 Fellars, P.J., 1991. The relationship between the ratio of degrees Brix to percent acid
- 573 and sensory flavor in grapefruit juice. Food Technol. 45(7): 68-75.
- 574 Flint M.L., 1991. Frost damage Integrated pest management for citrus., in: University of
- 575 California (Ed.), Integrated Pest Management Program for Citrus, Oackland, pp. 122-
- 576 Francis, F.J., 1995. Quality as influenced by colour. Food Quality and Preference, 6, 577 149-155.
- 578 García, J.M., Medina, R.J., Olías, J.M., 1998. Quality of strawberries automatically
- 579 packed in different plastic films. J. Food Sci. 63, 1037-1041.
- 580 Goto, A., Araki, C., 1983. Chemical composition and internal anatomy of the gelated
- and granulated juice sacs of Sanbokan fruit. J. Jpn. Soc. Hort. Sci. 52, 316-324.
- 582 Guthrie, J.A., Walsh, K.B., Reid, D.J., Liebensberg, C.J., 2005. Assessment of internal
- 583 quality attributes of mandarin fruit 1. NIR calibration model development. Aust. J.
- 584 Agric. Res. 56, 405-416.
- 585 Guthrie, J.A., Liebensberg, C.J., Walsh, K.B. 2006. NIR model development and
- 586 robustness in prediction of melon fruit total soluble solids. Aust. J. Agric. Res. 57, 411-
- 587 418.
- 588 Hernández, N., Lurol, S., Roger, J.M., Bellon Maurel, V. 2003. Robustness of models
- 589 based on NIR spectra for sugar content prediction in apples. J. NIR Spectrosc. 11, 97-
- 590 103.

591	Hernández, A, He, Y., García, A., 2006. Non-destructive measurement of acidity,	
592	soluble solids and firmness of Satsuma mandarin using vis/NIR-spectroscopy	
593	techniques. J. Food Eng. 77, 313-319.	
594	Ito, H. Morimoto, S. Yamuchi, R. Ippoushi, K. Azuma, K. Higashio H. 2002. Potential	
595	of near infrared spectroscopy for non destructive estimation of soluble solids in	
596	watermelons. Acta Hort. 588, 353-356.	
597	Iyo, C., Kawano, S. 2001. Predicting Brix values of stored apples using near infrared	
598	spectra. J. Japan. Soc. Hort. Sci., 1, 510-515.	
599	Kader, A.A., 1992. An overview, <u>Iin: Kader, A.A., (Eed), itor.</u> Postharvest technology	
600	of horticultural crops. Univ-ersity of California- Div. of agriculture and natural	
601	resources publications., Davies, Ppp. 15-20.	
602	Ladaniya, M.S. 2008. World fresh citrus tarde and quarantibe issues. In Citrus fruit:	
603	Biology, technology and evaluation. Elsevier. Pages 521-534.	
604	Kawano, S., Fujiwara, T., Iwamoto, M.J., 1993. Nondestructive determination of sugar	Formatted: Spanish (International Sort)
605	content in Satsuma mandarin using near infrared (NIR) transmittance. J. Jpn. Soc. Hort.	
606	<u>Sci. 62, 465–470.</u>	
607	Lehman-Salada, L., 1996. Instrument and operator effects on Apple firmness readings.	
608	Hort. Sci. 31, 994-997.	
609	Lu, R., Guyer, D.E., Beaudry, R.M., 2000. Determination of firmness and sugar content	
610	of apple using NIR diffuse reflectance. J. Texture Studies. 31, 615-630.	
611	Lu, R., 2001. Predicting firmness and, sugar content of sweet cherries using near	
612	infrared diffuse reflectance spectroscopy. Trans. ASAE. 44,1265-1271.	
613	McGlone, V.A., Kawano, S., 1998. Firmness, dry-matter and soluble-solids assessment	

614 of post-harvest kiwifruit by NIR spectroscopy. Postharvest Biol. Technol. 13, 131-134.

615	McGlone, V.A., Fraser, D.G., Jordan, R.B., Künnemeyer, R., 2003. Internal quality	
616	assessment of mandarin fruit by vis/NIR spectroscopy. J. NIR Spectrosc. 11, 323-332.	
617	Mercado-Silva, E., Benito-Bautista, P., García-Velasco, M.A., 1998. Fruit development,	
618	harvest index and ripening changes of guavas produced in Central Mexico. Postharvest	
619	Biol. Technol. 13, 143-150.	
620	Miyamoto, S., Serata, K., Uchigasaki, M., Aoyama, T., 1996. Relationship between the	Formatted: Spanish (International Sort)
621	transferred characteristics and internal qualities of fruit. ASAE ₁ paper <u>n°</u> 966055.	
622	Mowat, A.D., Poole, P.R., 1997. Use of visible-near infrared diffuse reflectance	
623	spectroscopy to discriminate between kiwifruit with properties altered by preharvest	
624	treatments. J. NIR Spectrosc. 5, 113-122.	
625	Muramatsu, N., Kiyohide, K., Tatsushi O., 1996. Relationship between texture and cell	
626	wall polysaccharides of fruit flesh in various species of citrus. HortSci. 31, 114-116.	
627	Nicolaï, B.M., Beullens, K., Bobelyn, E., Peirs, A., Saeys, W., Theron, K., Lammertyn,	
628	J. 2007. Non destructive measurement of fruit and vegetable quality by means of NIR	
629	spectroscopy: A review. Postharvest Biol. Technol. 46, 99-118.	
630	Olmo, M., Nadas, A., García, J.M., 2000. Non-destructive methods to evaluate maturity	
631	level of oranges. J. Food Sci. 65, 365-369.	
632	Onda, T., Tsuji, M., Komiyama, Y., 1994. Possibility of non-destructive determination	Formatted: Spanish (International Sort)
633	of sugar content, acidity and hardness of plum fruit by near-infrared spectroscopy. J.	
634	Japan. Soc. Food Sci. Tech. 41, 909-912.	
635	Peiris, K.H.S., Dull, GG., Leffler, R.G. Kays, S.J., 1997. Non-destructive determination	
636	of soluble solids content of peach by near infrared spectroscopy-, Hin: NRAES (Ed).	
637	Sensors for Non-destructive Testing International Conference and Tour, 18-21 February	
638	1997 Orlando Florida- pp. 77-87	

- 639 Peiris, K.H.S. Dull, G.G., Leffler, R.G. Kays, S.J., 1999. Spatial variability of soluble 640 solids or dry matter content within individual fruits, bulbs, or tubers: implications for the development and use of NIR spectrometric techniques. HortScience, 34, 114-118. 641 642 Pérez D, Sánchez MT, Paz P, Soriano MA, Guerrero JE, Garrido A., 2009. Non-643 destructive determination of quality parameters in nectarines during on-tree ripening 644 and postharvest storage. Postharvest Biol. Technol. 52, 180-188. 645 Polderdijk, J.J., Tijskens, L.M.M., Robberts, J.E., Van der Valk, H.C.P., 1993. 646 Predictive models of keeping quality of tomatoes. Postharvest Biol. Technol. 2, 179-647 185. 648 Poole, N.D., Martínez-Carrasco, L., Vidal, F., 2006. Quality perceptions under evolving 649 information conditions: Implications for diet, health and consumer satisfaction. Food 650 Policy. 32, 175-188. 651 Riquelme, M.T., 2008. Transmision óptica e imagen en visible e infrarrojo en frutas.
- 652 Ensayo de equipos comerciales. Tesis Doctoral. ETSIA, Universidad Politécnica de653 Madrid.
- Schatzki, T.F., Young, R., Le, L.C., Haff, R.P., Can, I., Toyofuku, N., 1997. Defect
 detection in apples by means of X-ray imaging. In: NRAES (Ed.) Sensors for Nondestructive Testing International Conference and Tour, 17-21 February 1997, Orlando,
 Florida.
- Schmilovitch, Z., Mizrach, A. Hoffman, A., Egozi, H., Fuchs, Y., 2000. Determination
 of mango physiological indices by near-infrared spectrometry. Postharvest Biol.
 Technol. 19, 245-252.

Formatted: Spanish (International Sort)

- 661 Shiina, T., Ijiri, T., Matsuda, I., Sato, T., Kawano, S., Ohoshiro, N., 1993.
 662 Determination of brix value and acidity in pineapple fruits by near infrared
 663 spectroscopy. Acta Hort., 334, 261-272.
- 664 Slaughter, D.C. 1992. Near infrared analysis of soluble solids in peaches. In ASAE
- 665 (Ed.) 1992 Summer Meeting, paper nº 92-7056, Charlotte, North Carolina.
- Sohn, M.R., Park, K.S., Cho, S.I., 2000. Near infra-red reflectance spectroscopy for
 non-invasive measuring internal quality of apple fruit. Near Infr. Anal. 1, 27-30.
- 668 Steinmetz, V., Crochon, M., Maurel, B.V., Fernández, J.L.G., Elorza, P.B., Taylor,
- M.A., Rabe, E., Jacobs, G., Dodd, M.C., 1995. Effect of harvest maturity on pectic
 substances, internal conductivity, soluble solids and gel breakdown in cold-stored
- 671 'Songold' plums. Postharvest Biol. Technol. 5, 285-294.
- 672 Tsuchikawa, S., Sakai, E., Inoue, K. 2003. Application of Time of flight Near infrared
 673 Spectroscopy to detect sugar and acid content in Satsuma mandarin. J. Amer. Soc. Hort.
- 674 Sci. 128, 391 396.
- Upchurch, B.L., Throop, J.A., 1994. Effects of storage duration on detecting watercore
 absorption to density and watercore content in apples. Transations of the ASAE, 37,
 873-877.
- 678 Van Kooten O., Schouten, R.E., Tijskens, L.M.M., 1997. Predicting shelf life of
- 679 cucmbers (*Cucumis sativus* L.) by measuring colour and photosyntesis₄₇ Lin: NRAES
- 680 (Ed.)₂ Sensors for Non-destructive Testing International Conference and Tour, 18 21
- 681 February 1997, Orlando, Florida, pp. 45-55.
- Verstreken, L., 1996. Sensors for fruit firmness comparison and fusion. Journal of
 Agricultural Engineering Research, 64, 15-28.

684	Watkins, C.B., 2008. Postharvest Ripening Regulation and Innovation in Storage
685	Technology. Acta Hort., 796, 51-58.
686	Williams, P., Sobering, D., 1996. How do we do it: a brief summary of the methods we
687	use in developing near infrared calibrations, In Davies, A.M.C. and Williams, P.
688	(Eds.), Near infrared spectroscopy: the future waves. NIR Publications Chichester.
689	<u>185-188.</u>
690	Wold, S., Martens, H., Wold, H. 1983. The multivariate calibration method in chemistry
691	solved by the PLS method, in: A. Ruhe, B. Kagström (Eds.), Proceedings on the
692	Conference on Matrix Pencils, Lecture Notes in Mathematics, Springer-Verlag,
693	Heidelberg, pp. 286–293.
694	Zude, M., Herold, B., Roger, J.M., Bellon Maurel, V., Landahl, S. 2006. Non-
695	destructive tests on the prediction of apple fruit flesh firmness and soluble solids content
696	on tree and in shelf life. J. Food Eng. 77, 254–260.
697	
698	(Abbot et al. 1995; Miyamoto et al. 1996), (Steinmetz et al. 1996), (DeEll, 1996;
699	models Van Kooten at al. 1997), (Schatzki et al. 1997 (Armstrong et al. 1997)
700	(Upchurch and Throop 1994), (Kader, 1992; Polderdijk et al., 1993; Blankenship et al.,
701	1997) Olmo et al. (2000)
702	
703	
704	Figure captions
705	Figure 1. A) Labspec, contact probe. B) Luminar 5030, hand held unit
706	Figure 21. A) Labspec, B) Luminar 5030. Examples of absorbance spectra from the
707	same five oranges

ī.

Figure 32. External validation plots V1.
A) Labspec, B) Luminar 5030. RPD values at the lower right corner.

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	Sanguinelli			Valencia			Salustiana			Navelate			Total		
	Ν	σ	\overline{X}	Ν	σ	\overline{X}	Ν	σ	\overline{X}	Ν	σ	\overline{X}	Ν	σ	\overline{X}
SSC	250	0.84	8.53	50	1.04	8.16	44	0.84	10.94	52	0.72	11.86	396	1.56	9.19
А	250	0.15	3.14	50	0.13	3.17	44	0.12	3.87	52	0.13	3.52	396	0.28	3.28
TA	250	0.19	0.78	50	0.21	1.48	44	0.09	0.70	52	0.16	1.06	396	3.12	9.38
MI	250	2.45	11.35	50	0.63	5.56	44	2.30	15.85	52	1.79	11.44	396	3.36	11.14
F	250	1.96	5.79	50	1.39	7.67	44	1.06	5.76	52	1.43	7.44	396	1.91	6.24
JV	250	7.25	31.88	50	9.40	48.58	44	6.29	76.80	52	7.87	74.27	396	19.80	44.55
FW	250	20.71	81.52	50	23.98	124.05	44	9.41	214.60	52	19.22	290.19	396	77.88	129.08
JV/FW	250	0.04	0.39	50	0.02	0.39	44	0.02	0.36	52	0.03	0.26	396	0.06	0.37
RW	250	14.16	43.62	50	14.70	72.09	44	6.62	128.15	52	17.59	209.72	396	59.29	78.42
FCI	250	8.62	21.05	50	8.40	15.07	44	6.08	26.48	52	3.54	18.51	396	8.35	20.56
JCI	250	88.35	62.97	50	24.52	-71.05	44	14.80	-35.62	52	30.18	-72.46	396	93.90	17.31

Table 1. Statistical data of the sets of samples and orange varieties.

SSC, soluble solids content (%); A, Acidity (pH); TA, titratable acidity (g/L citric acid); MI, maturity index; F, flesh firmness (N); JV, juice volume (mL); FW, fruit weight (g); RW, rind weight (g); FCI, fruit colour index; JCI, juice colour index; σ , standard deviation; \overline{X} , mean.

Parameter	Samples	Range	σ	\overline{X}
SSC	76	5.5-12.4	1.59	9.08
А	76	1.1-3.9	0.28	3.27
TA	75	0.6-2.6	0.29	0.91
MI	75	4.0-16.7	3.22	10.85
F	76	1.5-11.7	1.93	5.96
JV	74	18.0-88.0	20.58	45.00
FW	75	47.4-311.4	79.19	127.08
JV/FW	76	0.2-0.5	0.06	0.37
RW	76	18.5-228.5	60.0	76.5
FCI	72	8.7-31.0	4.64	20.54
JCI	76	-141.9-223.4	87.40	16.40

Table 2. Statistical data of the sets of samples used in V1.

SSC, soluble solids content (%); A, Acidity (pH); TA, titratable acidity (g/L citric acid); MI, maturity index; F, flesh firmness (N); JV, juice volume (mL); FW, fruit weight (g); RW, rind weight (g); FCI, fruit colour index; JCI, juice colour index; σ , standard Deviation; \overline{X} , mean.

			Val	idation	l				
Parameter	Treatment	PCs	Outl.	RMSECV	R	R _{CV}	RMSEP	RPD	Т
SSC	MN	10	0	0.60	0.92	0.91	0.74	2.13	0.719
А	MN, MSC	10	4	0.12	0.90	0.88	0.15	1.85	0.714
ТА	MN	10	0	0.16	0.86	0.83	0.17	1.69	0.978
MI	MN	8	0	1.81	0.85	0.81	1.92	1.67	0.765
FW	MN, MSC	10	2	19.89	0.97	0.96	16.52	4.76	0.716
F	MN	7	3	1.18	0.76	0.72	1.05	1.85	0.356
JV	MN	9	3	7.39	0.92	0.91	7.05	2.94	0.511
JV/FW	MN, MSC	9	3	0.03	0.89	0.87	0.04	1.61	0.327
RW	MN	10	2	14.61	0.97	0.96	12.98	4.54	0.349
FCI	MN, MSC	8	0	3.45	0.90	0.87	2.65	1.75	0.055
JCI	MN, MSC	10	0	1.59	0.86	0.83	66.78	1.31	0.286

Table 3. Labspec. Statistics of calibrations and validations. Wavelength 500-2300 nm.

SSC, soluble solids content (%); Acidity (pH); TA, titratable acidity (g/L citric acid); MI, maturity index; Flesh firmness (N); JV, juice volume (mL); FW, fruit weight (g); Rind weight (g); MN, mean normalisation; MSC, multiplicative scatter correction; Outl., outliers; RMSEC, root mean square error of calibration; R, coefficient of calibration; R_{CV}, coefficient of cross validation; RMSEP, root mean square error of prediction; RPD, residual predictive deviation; T, **p* value from paired samples test.

		Validation							
Parameter	Treatment	PCs	Outl.	RMSECV	R	R _{CV}	RMSEP	RPD	Т
SSC	MN, MSC	10	0	0.67	0.91	0.89	0.68	2.33	0.273
А	MN, MSC	10	1	0.15	0.84	0.81	0.16	1.75	0.931
ТА	MN	10	2	0.18	0.80	0.77	0.19	1.54	0.321
MI	MN, MSC	9	1	2.06	0.79	0.74	1.96	1.64	0.313
FW	none	9	3	22.69	0.96	0.95	26.26	3.03	0.414
F	MSC	10	0	1.27	0.71	0.66	1.39	1.39	0.280
JV	MN, MSC	9	2	7.97	0.91	0.91	8.00	2.56	0.977
JV/FW	MSC	9	1	0.03	0.83	0.80	0.04	1.67	0.148
RW	none	9	2	17.07	0.96	0.95	18.86	3.23	0.249

Table 4. Luminar 5030. Statistics of calibrations and validations. Wavelengths 1100-2300 nm.

SSC, soluble solids content (%); Acidity (pH); TA, titratable acidity (g/L citric acid); MI, maturity index; F, flesh firmness (N); JV, juice volume (mL); FW, fruit weight (g); RW, Rind weight (g). MN, mean normalisation. MSC, multiplicative scatter correction; Outl., outliers; RMSEC, root mean square error of calibration; R, coefficient of calibration; R_{CV}, coefficient of cross validation; RMSEP, root mean square error of prediction; RPD, residual predictive deviation; T, **p* value from paired samples test.

	Labsp	ec	Luminar 5	5030
Parameter	RMSEP	RPD	RMSEP	RPD
SSC	0.87	2.21	1.12	1.03
А	0.13	1.05	0.40	0.80
ТА	2.47	1.26	2.07	1.07
MI	1.54	1.75	2.57	1.26
FW	43.51	1.11	32.63	0.75
F	1.82	1.20	1.53	1.03
JV	8.38	1.10	12.13	0.84
JV/FW	0.04	1.18	0.05	1.10
RW	16.07	1.11	14.71	0.73
FCI	6.48	1.64		
JCI	55.69	0.67		

Table 5. Validation V2. N = 50. Wavelength 500-2300 nm.

SSC, soluble solids content (%); A, Acidity (pH); TA, titratable acidity (g/L citric acid); MI, maturity index; Flesh firmness (N); JV, juice volume (mL); FW, fruit weight (g); Rind weight (g); RMSEP, root mean square error of prediction; RPD, residual predictive deviation; N, number of samples of the validation set.



