A new internal quality index for mango and its prediction by external visible and near-infrared reflection spectroscopy

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ABSTRACT

A non-destructive method based on external visible and near-infrared reflection spectroscopy for determining the internal quality of intact mango cv. ‘Osteen’ was investigated. An internal quality index, well correlated with the ripening index of the samples, was developed based on the combination of a biochemical property (total soluble solids) and physical properties (firmness and flesh colour) of mango samples. The diffuse reflectance spectra of the samples were recorded and used to predict the internal quality and the ripening index. These spectra were obtained using different spectroscopic external measurement sensors involving a spectrometer, capable of measuring in different spectral ranges (600-1100 nm and 900-1750 nm), and also a spectrocolorimeter that measured in the visible range (400-700 nm). Three regression models were developed by partial least squares to establish the relationship between spectra and indices. Good results in the prediction of
internal quality of the samples were obtained using the full spectral range (R_p^2 = 0.833-0.879, RMSEP = 0.403-0.507 and RPD = 2.341-2.826) and some selected wavelengths (R_p^2 = 0.815-0.896, RMSEP = 0.403-0.537 and RPD = 2.060-2.905). The results obtained from this study revealed that external visible and near-infrared reflection spectroscopy can be used as a non-destructive method to determine the internal quality of mango cv. ‘Osteen’.

**Keywords**: reflection spectroscopy, fruit, quality, chemometrics, non-destructive.
1. INTRODUCTION

Spain is the main European producer of subtropical fruits, with approximately 1400 ha dedicated to mango (Galán & Farre, 2005). In particular, the south-west region has a large potential for the production of tropical and subtropical fruit, with a favourable year-round climate and infrequent frosts.

Mango fruit is sold in the market in quality categories. In the past, skin colour, fruit size and shape, freedom from defects and the absence of decay were the most common quality determinants, but nowadays other organoleptic characteristics related with internal and nutritional quality play an important role in the consumer's decision, as opposed to just appearance. The quality of mangoes changes almost daily and it is essential to correlate all the major quality parameters with one another in order to reveal the overall quality of the fruit (Jha et al., 2011).

In a climacteric fruit, such as mango, the fruit is not considered to be of desired eating quality at the time it initially becomes mature. It requires a ripening period before it achieves the taste and texture desired at the time of consumption. The ripening process is regulated by genetic and biochemical events that result in biochemical changes such as the biosynthesis of carotenoids (Mercadante & Rodriguez-Amaya, 1998), loss of ascorbic acid (Hernández et al., 2006), increase in total soluble solids (Padda et al., 2011); physical changes such as weight, size, shape, firmness and colour (Ornelas-Paz et al., 2008; Kienzle et al., 2011); and changes in aroma, nutritional content and flavour of the fruit (Giovannoni, 2004). Traditional determination of the internal quality of mango requires a destructive methodology using specialised equipment, procedures and trained personnel, which results in high analysis costs and does not allow
the whole production to be analysed (Torres et al., 2013). Nevertheless, new
technologies to monitor fruit quality changes during the postharvest handling
chain are rapidly being introduced, especially those based on non-destructive
assessment methods, recently reviewed by Jha et al. (2010) and Nicolaï et al.
(2014). These fast and non-destructive methods can help to provide decisive
parameters with which to obtain better quality mango products and to promote
consumption of mangoes with better health benefits (Ibarra-Garza et al., 2015).
Several non-destructive technologies have been widely explored to predict the
quality and maturity of mango, such as nuclear magnetic resonance (NMR) (Gil
et al., 2000), impact response (Padda et al., 2011; Wanitchang et al., 2011),
electronic nose (Lebrun et al., 2008; Zakaria et al., 2012), hyperspectral
analysis (Vélez-Rivera et al., 2014a), and near-infrared spectroscopy
(Saranwong et al., 2004). Conversely, some authors, such as Jha et al. (2005),
have included in their studies the full visible spectrum using spectroscopy in
intact mangoes, although studies using colour coordinates are more common,
such as Jha et al. (2007), Subedi et al. (2007) or Rungpichayapichet et al.
Schmilovitch et al. (2000) studied the feasibility of near-infrared spectroscopy
(NIRS) to determine the total soluble solids, firmness and acidity of mangoes
cv. ‘Tommy Atkins’ in relation to the maturity stage. Nagle et al. (2010)
developed a method to measure total soluble solids, total acidity and dry matter
in mango cv. ‘Chok Anan’ on the trees using NIRS. Theanjumpol et al. (2013)
studied the possibility of predicting six main chemical substances found in
mango fruit cv. ‘Keitt’ and cv. ‘Nam Dok Mai Si Thong’, which are important in
Thailand, such as glucose, sucrose, citric acid, malic acid, starch and cellulose.
They used VIS/NIR spectrometry but decided not to use the visible information to avoid the influence of colour pigments. Jha et al. (2013) studied the properties of different mangoes that are important for the Indian production using NIRS in the 1200–2200 nm range to measure properties that make it possible to predict the maturity stage. They were able to predict the sweetness of the mangoes from measurements of total soluble solids and pH (Jha et al., 2012) or to determine a maturity index based on the estimations of total soluble solids, dry matter and total acidity that was compared with destructive analysis and sensory panels, and corrected using a constant that depended on the cultivar (Jha et al., 2014). Watanawan et al. (2014) studied the 700–1100 nm region in an attempt to correlate total soluble solids, total acidity and dry matter of mango cv. ‘Namdokmai’ with maturity in order to predict the optimum harvesting time. They found good correlations among NIRS values and firmness and dry matter content at harvest, and predicted TSS with very high accuracy, although they consider that their study needs to be revised in order to reduce the heterogeneity in fruit maturity and increase the outturn quality.

The main problem of using NIRS to assess fruit quality is the robustness of the calibration model (Rungpichayapichet et al., 2016). Additionally, fruit cultivar, size, and harvest season also play an important role in the robustness of NIRS models (Bobelyn et al., 2010). In this study, a non-destructive method based on visible and near-infrared spectroscopy was investigated to determine the internal quality of mango cv. ‘Osteen’ during ripening because this is the main variety of mango grown in Spain. This variety could be included in the group of late-ripening mangoes, with higher weights and prices than other varieties of the same fruit. For this reason, this variety is considered to be optimal for export.
owing to its late maturing characteristics and final relatively low weight loss (Siller-Cepeda et al., 2009).

Hence, the aims of this research were (a) to determine an internal quality index for mangoes, based on their main biochemical (total soluble solids) and physical properties (firmness and flesh colour), avoiding the titratable acidity analysis, because it is a laborious and slow analysis that generates waste, (b) to apply it to mango cv. ‘Osteen’, and (c) to develop statistical models based on Partial Least Squares (PLS) to predict the internal quality of the samples through the analysis of external VIS-NIR spectral data.

2. MATERIALS AND METHODS

2.1. Experimental procedure

A batch of 140 unripe mangoes (Mangifera indica L., cv ‘Osteen’) were obtained from plantations in Málaga (Spain). The fruit selected were free of external damage or diseases, showing a uniform shape and size. All mangoes were washed and dried to completely remove any water from the surface and then were marked on each side. All sets were ripened in a storage chamber at 18.0 ± 2.1 °C and 67.6 ± 3.3% RH. Sets of twenty mangoes were randomly collected and analysed every two-three days until reaching senescence (sixteen days).

The visible and near-infrared spectra of the external skin of each mango were measured on the centre of one cheek and two points on the other cheek (Figure 1) on each day of storage. After the measurements, the physical and biochemical properties were analysed.
2.2. Visible and near-infrared spectra collection

The spectral characteristics of the external skin of the intact mangoes were measured in the visible and in the short and medium near-infrared range using a conventional spectrocolorimeter and a VIS-NIR spectrometer.

The external visible spectra of mango samples between 400 and 700 nm, every 10 nm, were measured using a spectrocolorimeter (CM-700d, Minolta Co., Tokyo, Japan). All the measurements were performed by placing the spectrocolorimeter directly onto the skin of the fruit.

The visible-near infrared and near-infrared spectra of mango samples were collected alternately in reflectance mode using a multichannel spectrometer platform (AVS-DESKTOP-USB2, Avantes BV, The Netherlands) equipped with two detectors (Figure 2). The first detector (AvaSpec-ULS2048 StarLine, Avantes BV, The Netherlands) included a 2048-pixel charge-coupled device (CCD) sensor (SONY ILX554, SONY Corp., Japan), 50 µm entrance slit and a
600 lines/mm diffraction grating covering the VIS-NIR range from 600 nm to 1100 nm with a spectral FWHM (full width at half maximum) resolution of 1.15 nm. The spectral sampling interval was 0.255 nm. The second detector (AvaSpec-NIR256-1.7 NIRLine, Avantes BV, The Netherlands) was equipped with a 256 pixel non-cooled InGaAs (Indium Gallium Arsenide) sensor (Hamamatsu 92xx, Hamamatsu Photonics K.K., Japan), a 100 µm entrance slit and a 200 lines/mm diffraction grating covering the NIR range of 900 nm to 1750 nm and a spectral FWHM resolution of 12 nm. The spectral sampling interval was 3.535 nm. A Y-shaped fibre-optic reflectance probe (FCR-7IR200-2-45-ME, Avantes BV, The Netherlands) was configured with an illumination leg which connects the fibre coupled to a stabilised 10 W tungsten halogen light source (AvaLight-HAL-S, Avantes BV, The Netherlands). The light source ensures a permanent light intensity over the whole measurement range. A holder was used to position the sample properly over the probe and the reflectance probe delivered the light to the sample and collected the reflectance from the sample, which was carried by the fibre cable to the spectrometer in use. The reflectance probe, consisting of seven fibres with a diameter of 200 µm, delivered the light to the sample through a bundle of six fibres. The probe tip was designed to provide reflectance measurements at an angle of 45° so as to minimise specular reflectance from the surface of the fruit. The calibration was performed using a 99% reflective white reference tile (WS-2, Avantes BV, The Netherlands) so that the maximum reflectance value over the range of wavelengths was around 90% of saturation.
Prior to spectral measurements, the temperature of the mangoes was stabilised at 24 ± 1 °C. Measurements were taken at three longitudinal points over the surface of the fruit and mean values of the spectra were used for the analysis. A personal computer equipped with commercial software (AvaSoft version 7.2, Avantes, Inc.) was used to control both detectors and to acquire the spectra. The signals were pre-processed using AvaSoft software. The integration time was set to 90 ms for the detector sensitive in the VIS-NIR region and to 700 ms for the detector sensitive in the NIR region. For both detectors, each spectrum was obtained as the average of five scans to reduce the thermal noise of the detector (Nicolaï et al., 2007). The average reflectance measurements of each sample (S) were then converted into relative reflectance values (R) with respect to the white reference using dark reflectance values (D) and the reflectance values of the white reference (W), as shown in equation (1):

\[ R = \frac{S-D}{W-D} \] (1)
The dark spectrum was obtained by turning off the light source and completely covering the tip of the reflectance probe.

2.3 Physical and biochemical analysis

The physical properties analysed were firmness, peel colour and flesh colour of the mangoes. The firmness, in Newtons, was analysed through a puncture test by using a universal test machine (TextureAnalyser-XT2, Stable MicroSystems (SMS) Haslemere, England). The tests were performed in triplicate in the axial direction at three locations in the equatorial section (Figure 1 (b)) with a punch with a diameter of 6 mm (P/15ANAMEsignature) until a relative deformation of 30%, at a speed of 1 mm/s.

CIE (Internationale de l'éclairage) colour values of Luminosity ($L^*$), chromaticity ($C_{ab}^*$) and hue angle ($h_{ab}^*$) for each fruit on both peel (external colour) and flesh (internal colour) were determined using the spectrocolorimeter. The standard illuminant D65 and the 10° standard observer were used for all colour measurements in the study. The colour values were averaged from three different measurements taken at three points on the fruit in order to have representative values. The biochemical properties analysed were the total soluble solids (TSS) and the titratable acidity (TA). TSS content was determined by refractometry with a digital refractometer (set RFM330+, VWR International Eurolab S.L., Barcelona, Spain) at 20 °C and was determined based on the percentage of soluble solids. The analysis of TA was performed with an automatic titrator (CRISON, pH-burette 24, Barcelona, Spain) with 0.5 N NaOH until a pH of 8.1 (UNE34211:1981), using 15 g of crushed mango which was
diluted in 60 mL of distilled water. The TA was determined based on the
percentage of citric acid that was calculated using equation 2.

\[
\text{Titratable acidity [g citric acid/100 g of sample]} = \frac{(A \times B \times C/D) \times 100}{E} \quad (2)
\]

where \(A\) is the volume of NaOH consumed in the titration (in L), \(B\) is the
normality of NaOH (0.5 N), \(C\) is the molecular weight of citric acid (192.1 g·mol\(^{-1}\)), \(D\) is the weight of the sample (15 g) and \(E\) is the valence of citric acid (\(E = 3\)).

Two indices, a ripening index (RPI) and an internal quality index (IQI) were
calculated by equations 3 and 4. The RPI was described previously by
Vásquez-Caicedo \textit{et al.} (2005) and Vélez-Rivera \textit{et al.} (2014b). However,
titratable acidity analysis is complex, laborious, slow and generates waste.
Furthermore, the colour has previously been proved to be a quality indicator of
mango (Jha \textit{et al.}, 2006a and 2006b). In Jha \textit{et al.} (2007) colour parameters
were highly correlated with TSS through the creation of several models based
on the CIEL*a*b* coordinates. From these studies, the IQI was calculated
combining TSS, firmness, and flesh colour. These parameters have been used
in packing houses to measure the quality of mangoes. They require less time,
less pre-treatment of the sample and lower costs.

\[
RPI = \ln(100 \cdot F \cdot TA \cdot TSS^{-1}) \quad (3)
\]

\[
IQI = \ln\left(100 \cdot F \cdot L^* \cdot h_{ab^*} \cdot TSS^{-1} \cdot C_{ab^*}^{-1}\right) \quad (4)
\]
where F is firmness (Newtons), TA is titratable acidity (%), TSS is total soluble solids (%) and $L^*$, $h_{ab}^*$ and $C_{ab}^*$ are the colour attributes of the flesh colour.

2.4. Statistical Analysis

The spectroscopic data and both indices were organised into three different matrices: the first matrix for the visible spectra (400-700 nm), the second matrix for the VIS-NIR spectra (600-1100 nm) and the third matrix for the NIR spectra (900-1750 nm). In all the matrices, the rows represent the number of samples (#N = 140 samples) and the columns represent the number of variables (X-variables and Y-variables). The X-variables, or predictors, were the different spectra and the Y-variables, or responses, were the two variables provided by RPI and IQI. All the matrices were analysed using The Unscrambler Version 9.7 software package (CAMO Software AS, Oslo, Norway). First, all the spectral data were pre-processed. The X-variables were transformed to apparent absorbance (log (1/R)) values to obtain linear correlations of the NIR values with the concentration of the estimated constituents (Shao et al., 2007; Liu et al., 2010) and centred by subtracting their averages in order to ensure that all results will be interpretable in terms of variation around the mean. Due to the high resolution causing an increased occurrence of signal noise by its spectral range measurement, the VIS-NIR spectra were reduced using a reduction factor of 7. In order to reduce the influence of light scattering (Santos et al., 2013) and the baseline drift various pre-processing methods were applied to the spectra. Savitzky-Golay smoothing with a gap of three data points combined with extended multiplicative scatter correction (EMSC) were considered the best results for the VIS-NIR spectra, and those two pre-treatments and second
derivative with Gap-Segment (2.3) were the best results for the NIR spectra.

After the pre-processing steps, the X-variables in the matrices were 31, 285 and 242 for the visible spectrum, VIS-NIR spectrum and NIR spectrum, respectively.

Secondly, each set was divided randomly into two groups, a calibration set (75% of the samples) and a prediction set (25% of the samples). Partial least squares regression (PLS) was applied to the matrix to construct separate calibration models for each ripening index and each spectrum with segments of 20 objects (Næs et al., 2004) and it was evaluated by means of a cross validation methodology. PLS defines the latent variables (principal components) based on the covariance between the independent and dependent variables, the advantage of PLS regression being its ability to analyse data with many, noisy, collinear, and even incomplete variables in both X and Y (Næs et al., 2004). This technique has usually been used in multivariate calibration in fruit applications (Liu et al., 2010) and allows obtaining the best results when linear relations between spectra and properties of samples exist (Li et al., 2010). Liu et al. (2008) used the MLR technique based on the regression of the discrete parts of the spectra and PLS based on the full spectrum; the results of the two techniques in their study appeared to be very similar.

In order to reduce the high dimensionality of the spectral data, to avoid the presence of noise or information that is not related to the quality characteristics of the mango, and to make the PLS models more robust, the most important wavelengths to predict both indices were selected (ElMasry et al., 2007; Talens et al., 2013). For each calibration model, the weighted regression coefficients resulting from the PLS models were used to select the important wavelengths. Regression coefficients show the weight of the contribution of each wavelength
to the calibration model and eliminate the spectral regions with less contribution.

Standardised spectral data were used to develop the PLS models to obtain the weighted regression coefficients.

The relative performance of the constructed models was assessed by the required number of latent variables (LVs), the coefficient of determination for calibration ($R^2_C$), the root mean square error of calibration (RMSEC) and the root mean square error of leave-one-out cross-validation (RMSECV). A model can be considered good when a low number of LVs are required and it has a low RMSEC and RMSECV and high $R^2_C$. The predictive ability of the models was evaluated using the coefficient of determination for prediction ($R^2_P$), the root mean square error of prediction (RMSEP) and the ratio of prediction to deviation (RPD=SD/RMSEP), where the SD was the standard deviation of the Y-variable in the prediction set. A value below 1.5 for the RPD indicates that the calibration is not usable. A value between 1.5 and 2.0 for the RPD reveals a possibility to distinguish between high and low values, while a value between 2.0 and 2.5 makes approximate quantitative predictions possible. For values between 2.5 and 3.0, and above 3.0, the prediction is considered to be good and excellent, respectively (Williams & Sobering, 1993; Saeys et al., 2005; Cozzolino et al., 2011). $R^2_C$ measured the performance of a multivariate calibration model and can be defined as the following Equations 5 and 6 (Yahaya et al., 2015):

\[
RMSEC = \sqrt{\frac{1}{n_C} \sum_{i=1}^{n_C} (\hat{y}_i - y_i)^2}
\]  

\[
RMSECV \lor RMSEP = \sqrt{\frac{1}{n_P} \sum_{i=1}^{n_P} (\hat{y}_i - y_i)^2}
\]
where:

\( \hat{y}_i \) is the predicted value of the ith observation

\( y_i \) is the measured value of the ith observation

\( n_c \) is the number of observations in the calibration set

\( n_p \) is the number of observations in the validated set

### 3. RESULTS AND DISCUSSION

#### 3.1. Changes in mango quality during ripening

Table 1 shows the range (minimum and maximum values), mean and standard deviation of the quality parameters analysed in the mango samples (\( \#N = 140 \) samples).

<table>
<thead>
<tr>
<th>Mango</th>
<th>TSS (%)</th>
<th>TA (%)</th>
<th>Firmness (N)</th>
<th>( L^{* \text{ext}} )</th>
<th>( C^{* \text{ext}} )</th>
<th>( h^{* \text{ext}} )</th>
<th>( L^{* \text{Int}} )</th>
<th>( C^{* \text{Int}} )</th>
<th>( h^{* \text{Int}} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Min</td>
<td>5.8</td>
<td>0.07</td>
<td>11.8</td>
<td>32.0</td>
<td>34.5</td>
<td>14.0</td>
<td>56.3</td>
<td>8.6</td>
<td>70.4</td>
</tr>
<tr>
<td>MAX</td>
<td>20.7</td>
<td>0.98</td>
<td>124.3</td>
<td>56.5</td>
<td>65.0</td>
<td>94.2</td>
<td>81.8</td>
<td>46.8</td>
<td>86.1</td>
</tr>
<tr>
<td>Mean</td>
<td>13.4</td>
<td>0.47</td>
<td>76.1</td>
<td>43.1</td>
<td>53.0</td>
<td>47.2</td>
<td>72.7</td>
<td>25.2</td>
<td>78.7</td>
</tr>
<tr>
<td>Sdev</td>
<td>3.5</td>
<td>0.22</td>
<td>35.8</td>
<td>4.7</td>
<td>6.3</td>
<td>18.3</td>
<td>6.4</td>
<td>9.1</td>
<td>4.0</td>
</tr>
</tbody>
</table>

The firmness ranged from 124.3 to 11.8 N. The peel luminosity, peel chroma and peel hue ranged from 32, 34.5, 14 to 56.5, 65, 94.2, respectively, whereas flesh luminosity, flesh chroma and flesh hue ranged from 56.3, 8.6, 70.4 to 81.8, 46.8 and 86.1, respectively. The TSS and the TA ranged from 5.85 to 19.50% and 0.97 to 0.07%, respectively. Similar values were observed by other authors during the ripening process of mangoes, working with other mango varieties such as ‘Alphonso’ (Yashoda et al., 2007), ‘Tommy Atkins’ (Lucena et al., 2007), ‘Nam Dokmai’ and ‘Irwin’ (Fukuda et al., 2014).
Table 2 shows the Pearson correlation coefficients and was calculated to check for significant inter-correlations between the parameters analysed in mango samples. The results indicated that, in general, peel/external colour showed lower correlations with respect to the other physical and biochemical properties, whereas higher correlations were found between firmness, flesh/internal colour and the biochemical properties. Positive correlations were found between firmness and internal $L^*$ (0.93), internal $h_{ab^*}$ (0.88) and TA (0.63), and negative correlations were found between firmness and internal $C_{ab^*}$ (-0.78) and TSS (-0.79).

<table>
<thead>
<tr>
<th>TSS (%)</th>
<th>TA (%)</th>
<th>Firmness N</th>
<th>$L^*$ ext</th>
<th>$C^*$ ext</th>
<th>$h^*$ ext</th>
<th>$L^*$ Int</th>
<th>$C^*$ int</th>
<th>$h^*$ int</th>
<th>RPI</th>
<th>IQI</th>
</tr>
</thead>
<tbody>
<tr>
<td>TSS (%)</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>TA (%)</td>
<td>-0.36</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Firmness (N)</td>
<td>-0.79</td>
<td>0.63</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$L^*$ ext</td>
<td>0.16</td>
<td>-0.10</td>
<td>-0.23</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$C^*$ ext</td>
<td>0.64</td>
<td>-0.52</td>
<td>-0.78</td>
<td>0.15</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$h^*$ ext</td>
<td>-0.06</td>
<td>0.20</td>
<td>0.08</td>
<td>0.62</td>
<td>-0.21</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$L^*$ Int</td>
<td>-0.83</td>
<td>0.55</td>
<td>0.93</td>
<td>-0.22</td>
<td>-0.77</td>
<td>0.09</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$C^*$ int</td>
<td>0.67</td>
<td>-0.59</td>
<td>-0.75</td>
<td>0.46</td>
<td>0.58</td>
<td>-0.72</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$h^*$ Int</td>
<td>-0.81</td>
<td>0.54</td>
<td>0.88</td>
<td>-0.16</td>
<td>-0.84</td>
<td>0.22</td>
<td>0.94</td>
<td>-0.67</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>RPI</td>
<td>-0.76</td>
<td>0.82</td>
<td>0.92</td>
<td>-0.19</td>
<td>-0.72</td>
<td>0.12</td>
<td>0.89</td>
<td>-0.77</td>
<td>0.83</td>
<td>1</td>
</tr>
<tr>
<td>IQI</td>
<td>-0.87</td>
<td>0.63</td>
<td>0.95</td>
<td>-0.29</td>
<td>-0.75</td>
<td>0.02</td>
<td>0.94</td>
<td>-0.87</td>
<td>0.89</td>
<td>0.94</td>
</tr>
</tbody>
</table>

Figure 3 shows the changes in firmness, peel and flesh colour, TA and TSS of mangoes at different days of storage. As expected, firmness values of ‘Osteen’...
mangoes decrease constantly during ripening. At the beginning of the process, the firmness remained fairly constant, although a pronounced decrease in the firmness values was observed from eleven to sixteen days of storage, the loss of firmness on the last day of storage being around 75% of the firmness recorded at the beginning of the study. A similar behaviour has been reported for other mango varieties such as ‘Alphonso’ (Yashoda et al., 2005), ‘Ataulfo’ (Palafox-Carlos et al., 2012) or ‘Keitt’ (Ibarra-Garza et al., 2015). These changes can be attributed to different factors, such as the enzymatic activity (Prasanna et al., 2007; Yashoda et al., 2007) and/or the solubilisation, de-esterification, and de-polymerisation of the middle lamella, accompanied by an extensive loss of neutral sugars and galacturonic acid (Singh et al., 2013), which modify the structural integrity of the cell wall and middle lamella.
Figure 3. Mean and standard deviation of firmness, peel colour, flesh colour, TSS and TA of mangoes at different days of analysis.

The changes in peel/external and flesh/internal colour observed during the ripening of mangoes cv. ‘Osteen’ are also shown in Figures 3 and 4. Whereas flesh luminosity and flesh hue values decreased from 79° to 63° and from 83° to 74°, respectively, and flesh chroma values increased from 16 to 36, no clear changes in peel luminosity and peel hue values and small differences in peel chroma values could be observed, which is logical since the colour of the peel is heterogeneous and varies from one sample to another. In general, the peel colour changes were not uniform, indicating that peel colour is not an adequate quality parameter for cv. ‘Osteen’ mango cultivars.
However, flesh colour changes were uniform when fruit advances in ripening and can serve as an adequate quality parameter (Figures 3 and 4). The increase in the yellow-orange intensity of mango flesh can be associated with an increase in carotenoid content of mango fruit, as has been reported previously by other authors (Ornelas-Paz et al., 2008; Ibarra-Garza et al., 2015). This change is accompanied by a decrease in the $L^*$ value, although, despite the correlation, there is no evidence that the changes in the luminosity of the flesh ($L^*$) are actually due to the increase in carotenoids.

![Day of storage chart](image)

Figure 4. External appearance and flesh colour of mango cv. ‘Osteen’ at different days of storage.

Loss of firmness and changes in flesh colour correlate with the increase of the TSS ratio and decrease of TA (Figure 3). During ripening the TSS increased due to the conversion of starch into glucose and fructose, which are used as substrates during fruit respiration (Eskin et al., 2013), while the TA tends to
decrease due to the cell metabolisation of volatile organic acids and non-volatile constituents (Padda et al., 2011).

Taking into account the strong correlation found between the biochemical properties (TSS and TA) and the firmness and flesh colour (Table 2), two indices were calculated. The ripening index, RPI, involves the most essential physical and biochemical properties of the fruit linked with the sensory perception of the ripeness of the mangoes. The internal quality index, IQI, was calculated because it is a good indicator to assess changes in the mesocarp during the ripening of mangoes. In fact, firmness, total soluble solids and flesh/internal colour are the three parameters used in mango packing-lines to assess mango quality and stage of ripeness (Brecht et al., 2010), whereas the TA is more difficult and laborious to determine. Table 2 shows the Pearson correlation coefficient between the two indices, with higher positive correlations (0.94). Figure 4 shows the changes in the RPI and IQI indices calculated for the mangoes at different days of storage. In both cases, it can be observed that the values of the indices decreased during ripening. Based on previous studies working with RPI in mango cv. 'Manila' (Vélez-Rivera et al., 2014b) and comparing the values of this study, three ripeness phases were identified: unripe mangoes (values higher than 6, day two), intermediate-ripe mangoes (values between 6 and 4, days four to eleven) and over-ripe mangoes (values less than 4, days fourteen to sixteen), the intermediate-ripe mangoes being the mangoes with the best quality.
3.2. Analysis of visible and near-infrared spectra

When assessing ripening with a visible and/or near-infrared spectroscope, it is crucial to identify the spectral changes associated with pigment evolution and compositional changes. Typical apparent absorbance spectral of mangoes at different ripening stages for the visible region, the VIS-NIR and the NIR regions are shown in Figure 6.
Figure 6. Apparent absorbance spectra of mangoes at different ripening stages for the (a) visible region, (b) the VIS-NIR region and (c) the NIR region after pre-treatments.

All the spectra have a similar pattern with the main maxima located in the visible and near-infrared region, which showed the strong absorbance characteristics of the mangoes within the range of study. In the visible range, the light peaks around 400-500 nm are correlated with the carotenoid pigments (Lichtenthaler & Buschmann, 2001) and the peak around 640-700 nm illustrated the colour transition of mangoes correlated with the chlorophyll content that absorbs radiation in this region (Merzlyak et al., 2003). Similarly, Knee (1980) and Bodria et al. (2004) analysed apples and claimed that their reflectance minimum in the 670 nm to 680 nm range was strongly related to chlorophyll content. Therefore, the ripening process of the fruit, with changes in chlorophyll, carotenoid and anthocyanin contents, indicated the influence of pigment content and composition on the colouration of the entire spectral visible reflectance of the fruit (Yahaya et al., 2014; Omar, 2013). This view is supported by the study of Magwaza et al. (2012), who described the pattern of the absorption curves for Satsuma mandarin, which is similar to that for other fruit like mangoes and kiwis. On the other hand, the water peaks were recorded at around 950-1050 nm and 1350-1550 nm due to the second overtone of the OH stretching band (Bünning-Pfaue, 2003), and the variations at 1100-1250 nm are correlated with the sugar content (Osborne et al., 1993; Walsh et al., 2004). Figure 5a and 5b shows an increase in absorbance within the blue region during ripening, mainly linked to an increase in the carotenoids content, and a decreased
absorbance in the red region, mainly linked to a decrease in the chlorophyll content. Merzlyak et al. (2003) suggested that carotenoid synthesis is induced when chlorophyll degradation occurs during fruit ripening and senescence. Also during ripening, an increase in TSS is produced and could be mainly due to hydrolysis of starch into soluble sugars such as sucrose, glucose and fructose (Agravante et al., 1990; Cordenunsi & Lajolo, 1995).

3.3. Non-destructive prediction of mango quality

Multivariate analysis was performed in order to establish the quantitative relationship between the absorbance spectra and the internal quality of mango. The full range spectra for the three regions studied were used to establish calibration models based on PLS to explain RPI and IQI. The performance of the calibration models was optimised by internal cross-validation and then validated by external validation in an independent validation set. Table 3 and Table 4 show the results obtained for the calibration and cross-validation sets for the three models developed. Similar results on the calibration and cross-validation sets were obtained to predict RPI and IQI using the VIS, VIS-NIR or NIR detector. The models were very accurate with high $R_c^2$ (0.902-0.934) and $R_{CV}^2$ (0.831-0.903), while RMSEC (0.335-0.509) and RMSECV (0.395-0.546) were low. The models applied to the independent validation set were capable of predicting RPI with $R_p^2$ of 0.871, 0.902 and 0.845, and RMSEP of 0.520, 0.470 and 0.592, respectively, for the three spectral regions. On the other hand, the results achieved for the IQI were $R_p^2$ of 0.879, 0.877 and 0.833, and RMSEP of 0.464, 0.435 and 0.507, respectively, for the VIS, VIS-NIR and NIR detector. Although better results were obtained when visible information
was used, the models developed using the VIS/NIR and NIR spectra also presented high values of $R^2$ and low values of RMSEP.

Table 3. Results of the PLS models for the calibration and prediction of RPI in mango samples by using the full spectral range and the important wavelengths.

<table>
<thead>
<tr>
<th>DETECTOR</th>
<th>#W</th>
<th>#LV</th>
<th>$R^2$</th>
<th>RMSEC</th>
<th>$R^2_C$</th>
<th>RMSECV</th>
<th>$R^2_CV$</th>
<th>RMSEP</th>
<th>RPD</th>
</tr>
</thead>
<tbody>
<tr>
<td>VIS</td>
<td>31</td>
<td>6</td>
<td>0.907</td>
<td>0.415</td>
<td>0.886</td>
<td>0.463</td>
<td>0.871</td>
<td>0.520</td>
<td>2.916</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>4</td>
<td>0.893</td>
<td>0.445</td>
<td>0.882</td>
<td>0.471</td>
<td>0.871</td>
<td>0.520</td>
<td>2.827</td>
</tr>
<tr>
<td>VIS-NIR</td>
<td>285</td>
<td>8</td>
<td>0.934</td>
<td>0.335</td>
<td>0.902</td>
<td>0.412</td>
<td>0.902</td>
<td>0.470</td>
<td>2.767</td>
</tr>
<tr>
<td></td>
<td>6</td>
<td>6</td>
<td>0.847</td>
<td>0.509</td>
<td>0.827</td>
<td>0.546</td>
<td>0.795</td>
<td>0.548</td>
<td>2.373</td>
</tr>
<tr>
<td>NIR</td>
<td>242</td>
<td>10</td>
<td>0.922</td>
<td>0.364</td>
<td>0.868</td>
<td>0.478</td>
<td>0.845</td>
<td>0.592</td>
<td>2.340</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>5</td>
<td>0.853</td>
<td>0.499</td>
<td>0.830</td>
<td>0.542</td>
<td>0.831</td>
<td>0.613</td>
<td>2.259</td>
</tr>
</tbody>
</table>

Table 4. Results of the PLS models for the calibration and prediction of IQI in mango samples by using the full spectral range and the important wavelengths.

<table>
<thead>
<tr>
<th>DETECTOR</th>
<th>#W</th>
<th>#LV</th>
<th>$R^2$</th>
<th>RMSEC</th>
<th>$R^2_C$</th>
<th>RMSECV</th>
<th>$R^2_CV$</th>
<th>RMSEP</th>
<th>RPD</th>
</tr>
</thead>
<tbody>
<tr>
<td>VIS</td>
<td>31</td>
<td>6</td>
<td>0.916</td>
<td>0.363</td>
<td>0.903</td>
<td>0.395</td>
<td>0.879</td>
<td>0.464</td>
<td>2.826</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>4</td>
<td>0.881</td>
<td>0.433</td>
<td>0.871</td>
<td>0.455</td>
<td>0.838</td>
<td>0.537</td>
<td>2.522</td>
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<tr>
<td>VIS-NIR</td>
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<td>4</td>
<td>0.905</td>
<td>0.389</td>
<td>0.891</td>
<td>0.421</td>
<td>0.877</td>
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<td>2.691</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>5</td>
<td>0.827</td>
<td>0.525</td>
<td>0.796</td>
<td>0.575</td>
<td>0.896</td>
<td>0.403</td>
<td>2.905</td>
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<tr>
<td>NIR</td>
<td>242</td>
<td>10</td>
<td>0.902</td>
<td>0.394</td>
<td>0.831</td>
<td>0.523</td>
<td>0.833</td>
<td>0.507</td>
<td>2.341</td>
</tr>
<tr>
<td></td>
<td>7</td>
<td>5</td>
<td>0.841</td>
<td>0.503</td>
<td>0.820</td>
<td>0.540</td>
<td>0.815</td>
<td>0.531</td>
<td>2.060</td>
</tr>
</tbody>
</table>

The RPD values of the resulting models gave the relative predictive performance of the model more directly in comparison to either $R^2$ or RMSEP used alone. In this study, the RPD values obtained were 2.916, 2.767 and 2.340 for RPI and 2.826, 2.691 and 2.341 for IQI for the VIS, VIS-NIR and NIR detectors, respectively. In all cases they are high values indicating a greater ability of the models to accurately predict the internal ripeness of the mango in new samples.
Figure 7 shows the regression coefficient plots with the important wavelengths for RPI and IQI for each spectral range. These wavelengths corresponded to -H and -OH functional groups, which are related to carbohydrates (namely sugars and starches), organic acids and water (Rungpichayapichet et al., 2016).

After identifying the optimal wavelengths, the reduced sets of bands were used to build new PLS models using the absorbance at these particular wavelengths as independent variables, and the measured values of RPI and IQI as dependent variables. Figure 8 shows the efficiency of PLS models for this...
prediction, indicating that it is possible to use a reduced number of bands in the visible and near-infrared region to predict the internal quality of mango 'Osteen'.

**Figure 8.** Predicted versus measured values of RPI and IQI for the visible region (a), the visible-near infrared region (b) and near-infrared region (c).

The feasibility of VIS-NIR to predict RPI of mango was indicated by an $R_p^2$ between 0.795-0.871, RMSEP of 0.520-0.613 and RPD of 2.259-2.827 (Figure 8). The results corroborated other studies on the use of spectroscopy techniques to predict RPI, such as Mahayothee (2005), with $R^2$ of 0.8 and SEP of 0.9, or Runpichayapichet *et al.* (2016) with $R^2 > 0.8$ and SEP > 0.8 for 'Nam Dokmai' mangoes. Other indices that have been developed, for example, Jha *et al.* (2007), predicted the maturity index ($I_m$) using colour values with a
correlation of 0.92 and SEP of 10.72 for mango cv. Dashehari. Likewise, IQI has been identified as a quality indicator for mango cv. ‘Osteen’, and in this study $R_p^2$, RMSEP and RPD values were between 0.815-0.896, 0.403-0.537 and 2.060-2.905 respectively. As shown in Table 3 and Table 4, the PLS models created from the selected wavelengths reduced the number of latent variables while maintaining a similar performance to PLS models created with the full spectrum. Likewise, their calibration and prediction errors do not worsen and both indices remain the same range in mango samples.

4. CONCLUSIONS

The internal quality of intact mango ‘Osteen’ fruit has been assessed using external visible and near-infrared reflection spectroscopy. In order to assess the internal quality of the fruit, two indices have been used, the ripening index (RPI) and the internal quality index (IQI). Different spectroscopy systems were used to measure different spectral ranges (VIS, VIS-NIR and NIR) externally in reflectance mode. The partial least squares regression analysis showed a strong performance in predicting RPI and IQI for the VIS, VIS-NIR and NIR detectors using the full spectral range and the most important wavelengths. However, accuracy may be compromised when the measuring is being implemented on samples with different external geometry by two spectroscopy systems that are structured differently in terms of their optical-electronics configuration within the spectrometer or in their interfacing with the sample. Nevertheless, the results obtained from this study clearly reveal that external visible and near-infrared spectroscopy combined with chemometrics can be used for the non-destructive prediction of the internal quality of mango ‘Osteen’.
This technological development could even be integrated in continuous fruit packing lines as part of the quality assurance system.

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