Specialized Optical Fiber Sensor for Nondestructive Intrinsic Quality Measurement of *Averrhoa Carambola*

Ahmad Fairuz OMAR* and Mohd Zubir MATJAFRI

School of Physics, Universiti Sains Malaysia, 11800 Penang, Malaysia

*Corresponding author: Ahmad Fairuz OMAR      E-mail: thinker_academy@yahoo.com

Abstract: This paper presents an innovative and low-cost approach for nondestructive fruit quality analysis. The specialized optical fiber sensor developed and presented in this paper used a monochromatic wavelength, rather than a broad spectrum, to measure the intact carambola (star fruit) intrinsic quality, namely pH and firmness. The main objective of this research was to investigate the two optical fiber sensors used in this work, namely, the optical fiber red system (OF-RS) that operated with the peak sensitivity at 635 nm and the optical fiber near the infrared spectroscopy system (OF-NIRS) that operated with the peak sensitivity at 880 nm. Both systems showed good accuracy in the pH and firmness measurement of the intact carambola with the correlation coefficient $R$ over 0.75, and the measurement results were comparable with those of the commercial spectrometer. The best measurement results were obtained using OF-RS (pH: $R = 0.876$; the root mean square error ($RMSE$) = 0.211 pH; firmness: $R = 0.872$; $RMSE = 0.909$ kgf).

Keywords: Carambola, firmness, optical fiber sensor, pH, quality

1. Introduction – transformation in fruit quality assessment

For many decades, various efforts for fruit quality assessment have been theoretically and practically implemented. These efforts include the quality specification and instrumental measurement of the fruit quality [1]. A quality assessment system is required to ensure that the consumer purchases a fruit with the ideal quality [2]. Through continuous research, much progress has been achieved, and the agriculture industry has experienced transition in defining the term “quality” as well as the methodology of its interpretation. The current standard used in classifying the fruit quality is based on the size, shape, and presence and size of external damages [3]. Consequently, fruit properties that relate to sensory benefits, such as the chemical composition and texture, are not considered [4, 5]. Realizing the importance of these parameters on consumers’ acceptance, growers and distributors are developing company specifications that include relevant intrinsic properties acceptable to the consumers, such as the firmness and sugar and acid contents [3]. Currently, measuring these parameters usually requires destructive procedures and much labor and time. For that reason, a simpler, faster, and highly accurate measurement method is much preferred [6]. Chen, Sun [7] and Gao et al. [8] wrote a brief review listing the common methods and instruments that used for fruit quality assessment,
such as the optical technique, machine vision technique, electrical technique, sonic vibration, nuclear magnetic resonance, electronic noses, and computed tomography. The optical method, particularly spectroscopy, is one of the most practical and most successful techniques for nondestructive measurement of the fruit quality due to its high-sensitivity detection, good adaptability, lightweight equipment, and flexible usage, and it is not harmful to people [7, 8]; it has many advantages compared with the classical chemical and physical analytical methods [9].

Visible and near infrared (NIR) spectroscopy has been applied widely in the food industry, including the control and measurement of fruit and vegetable qualities [9, 10]. For instance, Fan et al. [11] conducted experiments to determine the soluble solid content (SSC) and firmness of apples using visible and NIR spectroscopy. In other recent spectroscopy applications in fruit quality assessment, Camps and Christen [12] measured the SSC, total acidity, and firmness of apricots, Cao et al. [13] measured the SSC and pH of grapes, Valente et al. [14] measured the firmness of mangoes, and Shao et al. [15] measured the SSC and pH of peaches. Meanwhile, Magzawa et al. [16] wrote a comprehensive review on the application of spectroscopy for internal and external quality analysis of citrus fruits. On the other hand, Sun et al. [17] concluded that visible and NIR spectroscopy was an excellent technology to assess the internal qualities of watermelons/melons and other fruits and, hence, was projected to play an important role in the research of online nondestructive techniques.

Commonly, in nondestructive fruit quality evaluation through spectroscopy using diode-array spectrometers, the measurement of the intrinsic parameters is conducted through the acquisition of broad spectra, which range from the visible up to the NIR wavelengths. A specific range of wavelengths represents the absorbance bands of a specific biochemical composition or color changes in the fruits at different stages of ripeness. Visible wavelengths between 672 nm and 676 nm are related to the chlorophyll contents of the fruits [13, 18, 19]. The NIR wavelengths usually used in the fruit intrinsic quality measurement are 950 nm, 960 nm, 970 nm, and 975 nm, which are related to the water absorbance or O–H bands [19–24]. The 910 nm to 914 nm wavelengths are related to the SSC or the third overtone of the C–H stretch [6, 24–26], and the 910 nm to 925 nm and 986 nm to 995 nm wavelengths are related to the pH [19, 22, 27]. Certain spectral regions that do not contain valuable information about the chemical variations in the samples or contain irrelevant information such as the noise and background, which can worsen the measurement model, are neglected [28, 29]. The combination of the useful bands will result in better regression between the optical and actual values. Nonetheless, it is less economical and highly complicated when the measurement is performed only using a small number of selected wavelengths while neglecting the rest of the unused range of the measurement spectra. Inexpensive online sensors and instruments for nondestructive determination of the fruit intrinsic quality can be developed if the performance based on effective wavelengths could be close to or higher than whole spectra [13]. Chauchard et al. [30] claimed that, instead of relying on miniature holographic gratings or bandpass filters, such as the NIR imaging spectroscopy experiment by Tsuta et al [31] to measure the SSC in melons, using monochromatic light sources such as the light-emitting diode (LED) or laser diode to measure specific chemical components coupled to a silicon photodiode detector was another option. Chauchard et al. [30] stated that this was an ideal alternative because of the combination of the small size, low cost, and better robustness. Hence, this paper presents an innovative application of the optical fiber sensor that uses LEDs and photodetector for the measurement of the fruit intrinsic quality, namely, pH and firmness. At this point of the
research, experiments have been conducted specifically on carambola (star fruit) from the B10 cultivar. The main objective of this paper is to identify the capability of the monochromatic optical fiber sensor in obtaining measurement accuracy that is comparable with that generated by commercial spectrometers.

2. Materials and methods

2.1 Overview on the quality assessment of Carambola

In Malaysia, for marketing purposes, the B10 carambola is classified according to seven different maturity indexes, as listed in Table 1; this classification is based on the color attributes. The grading process is currently being done manually by the Malaysian Federal Agricultural Marketing Authority (FAMA) through human visual identification, which is slow and inconsistent relative to the grading precision [32]. Because of a significant increase in the volume of B10 carambola production from 8,719 tons in 2005 to 16,915 tons in 2009 and 11,820 tons from January to June 2010 [33], efforts have been made by researchers to apply the latest technology for automatic carambola classification. Kamil et al. [32], Abdullah et al. [34], Abdullah et al. [35], and Amirulah et al. [36] have applied the machine vision technique using a charge-coupled device camera and image processing to classify carambola according to the FAMA standard index (Table 1). In addition to proposing an innovative and low-cost instrumentation for fruit quality analysis, this paper also redefines the term “quality” in the carambola classification, currently based on the physical properties (i.e., color, size, and shape), into intrinsic properties (i.e., pH and firmness).

2.2 Determination of pH and firmness

All the samples have to be tested for their actual quality attributes using a well-established instrumentation available commercially. The first intrinsic quality measurement conducted on the carambola samples was the firmness test using the Wagner FT Fruit Tester penetrometer with a full-scale measurement of 14 kgf and a resolution of 0.1 kgf. The penetrometer was attached to the FT 516 tip [5/16” (8 mm) diameter]. The penetrometer was applied at 90° of the fruit surface. The measurement was repeated at all five sides of the carambola sample. The average of all measurements was used to represent the firmness value of the fruit sample. After the fruit has been tested for its firmness, it was then cut into small cubes and pressed using a garlic squeezer, and its juice was collected in a 100-ml beaker. The acidity of the carambola was measured using the ExStik pH meter (PH100) from Extech Instruments (Waltham, Massachusetts, USA) with the measurement range between 0.00 and 14.00 pH, resolution of 0.01 pH, and accuracy of ±0.01 pH. The pH meter was calibrated using a buffer solution with pH values of 7 and 4. Table 2 lists the characteristics of the carambola used in our research.

Table 2 Carambola samples used in this experiment.

| Sample       | Weight (g) | Acidity (pH) | Firmness (kgf) |
|--------------|------------|--------------|----------------|
| Carambola (B10) | 95.45–206.19 | 2.41–4.12    | 1.5–8.0        |

2.3 Optical fiber sensor

The innovative approach presented in this paper is through the application of the light source and a detector with the narrow emissive bandwidth and spectral sensitivity. The optical fiber red system
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(OF-RS) and optical fiber NIR system (OF-NIRS) have been developed to measure the carambola intrinsic quality. Similar to the typical reflectance/absorbance spectroscopy experimental setup, the OF-RS and OF-NIRS also comprise a light source and a detector system. The OF-RS operates with the peak responsivity at 635 nm, whereas the OF-NIRS operates with the peak responsivity at 880 nm. The peak sensitivity definition is based on the peak emission wavelength of the LED, which is also closely related to the peak response of the optical sensor. The current limitation in the development of specialized optical fiber sensor is that only the restricted peak emission wavelengths and bandwidth of LEDs are available in the market. The same scenario also is applied in the selection of the optical detector, where the peak sensitivity is commonly available only at a certain and limited range. Despite these constraints, the measurement of the carambola intrinsic quality has been successfully conducted using the OF-RS and OF-NIRS, and a breakthrough that introduces a cost-effective, noninvasive measurement of the fruit intrinsic quality is presented in details in this paper.

The maximum efficiency of the optical measurement system can be achieved by matching accurately the source and the detector, particularly on their spectral response [38]. Therefore, selecting the light source and detector with ideal optical characteristics between each other is important. In this work, the OF-RS adopted the red LED “HLMP-EG08-WZ000” (AlInGaP emitter), which has a peak emissive wavelength at 635 nm. The NIR LED “OPE5587” (high-speed GaAlAs infrared emitter) has a peak emission wavelength at 880 nm.

For the optical fiber sensor design, two types of light detectors were used. The TSLR257, which has a high-sensitivity color light-to-voltage converter with a peak sensitivity at 635 nm and 670 nm (highest peak), was integrated into the OF-NIRS. Both optical detectors used in this work came from a similar sensor family manufactured by TAOS Inc., which combined a photodiode and a transimpedance amplifier with a gain of 320 MΩ on a single CMOS IC. The TSLR257 detector was equipped with a color filter over the photodiode, whereas the TSL267 was equipped with a visible light blocking plastic side-looker package with an integral lens.

In the development of the OF-RS and OF-NIRS, a photodetector was connected to the instrumentation amplifier (INA122) with either five or ten voltage amplification gains. Five voltage amplification gains could be achieved by leaving both R_G pins on the INA122 open, whereas ten amplification gains could be obtained by connecting a 40-kΩ resistor between the R_G pins. However, for the experiment on carambola quality measurement, the amplification gain was set to five for the optimum result. The amplification result at the output port was connected to a 12-bit serial analog-to-digital converter (MCP3202). The signal processing and data display were controlled by the Basic Stamp 2pe (BS2pe) microcontroller. The Basic Stamp 2pe motherboard (dimension: 69.9 mm × 34.3 mm) was used for data processing, which was capable of displaying the final result (ADC output) onto a four-digit, seven-segment display. The complete system circuit design is shown in Fig.1. The circuit shown represents the OF-RS and OF-NIRS.

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![Fig. 1 Circuit diagram for the optical fiber system developed for carambola quality measurement.](image-url)
2.4 Jaz spectrometer

Although every LED and the photodetector are considered for their peak wavelengths, these components actually have broad spectral characteristics. Therefore, we must understand how the individual wavelength within the spectral range of the LED and the detector used in this research responds to the quantification of pH and firmness of the carambola. Hence, the Jaz spectrometer from Ocean Optics Inc. ($\lambda = 200$ nm and $1,100$ nm) and the tungsten halogen lamp with a spectral emission between $360$ nm and $2,000$ nm were used for comparative analysis. During the calibration, the measurement probe was located at $5$ cm from the surface of the standard white reference while during the measurement on the intact carambola, the probe was located right on the top of the sample as shown by Fig.2. The SpectraSuite software was then used to convert the raw spectra into reflectance as represented by (1). The value of reflectance at the selected individual wavelength ($\lambda$: $600$ nm – $700$ nm & $\lambda$: $830$ nm – $930$ nm) was then correlated with the value of carambola pH and firmness.

$$R_\lambda = \frac{S_\lambda - D_\lambda}{R_\lambda - D_\lambda} \times 100$$  \hspace{1cm} (1)

where,

$S_\lambda$ = the sample intensity at the wavelength $\lambda$;

$D_\lambda$ = the dark intensity at the wavelength $\lambda$;

$R_\lambda$ = the reference intensity at the wavelength $\lambda$.

2.5 Experimental setup

Figure 2 shows the experimental setup for the measurement of the carambola intrinsic quality through the optical fiber sensor (Fig.3) as well as the spectrometer via the interactance mode. The reflectance measurement mode did not produce any results in the measurement, which was most probably because of the glossy carambola surface that led to the high specular reflection and minimized the actual signal produced diffusely by the fruit. Furthermore, due to the limited penetration depth of the NIR radiation into the fruit, the reflectance spectra did not contain much information about the internal quality of the flesh [16, 39]. This problem was solved with the implementation of the interactance measurement mode, done by allowing the light to enter directly into the fruit, and the retrieval of the resultant light was made at a distance from the emission probe. In other words, the field of view of the detector was separated from the illuminated surface [40].

The accuracy of the regression models was calculated using the correlation coefficient, $R$, and root mean square of error, $RMSE$ (in pH for acidity and kgf for firmness) using (2) and (3).

The coefficient of determination is expressed as

$$R = \frac{\sqrt{n\sum xy - (\sum x)(\sum y)^2}}{\sqrt{n\sum x^2 - (\sum x)^2} \sqrt{n\sum y^2 - (\sum y)^2}}$$  \hspace{1cm} (2)

where $x$ is the independent variable, and $y$ is the dependent variable.

The root mean square of error is expressed as

$$RMSE = \sqrt{\frac{\sum (y - \bar{y})^2}{n-1}}$$  \hspace{1cm} (3)
where \( n \) is the number of data points, and \( Y \) is the predicted \( y \)-value.

3. Results and analysis

The first part of this section presents the results from the interactance measurement of the pH and firmness of the intact carambola using the Jaz spectrometer, whereas the second part elaborates on the result obtained using the optical fiber sensors. Figure 4 shows the relationship between the linear correlation coefficient and the individual wavelengths between 600 nm and 700 nm in quantifying the carambola pH and firmness. From the graph shown in Fig.4, for the pH measurement, the wavelengths between 655 nm and 685 nm produce a high correlation coefficient \( R \) over 0.85 with a peak response at 675 nm \((R = 0.876)\). For the firmness measurement, the wavelengths between 615 nm and 690 nm produce a high \( R \) (over 0.8) with a peak response at 660 nm \((R = 0.873)\). For the measurement of both parameters, a drastic drop in \( R \) is observed for the wavelength beyond 690 nm. Overall, similarities are observed in the response curve between the pH and firmness measurement. The firmness measurement shows a better response compared with that of the pH for wavelengths between 600 nm and 660 nm, whereas for wavelengths between 665 nm and 700 nm, the pH measurement shows a better response compared with that of the firmness. The OF-RS operated within this wavelength range, with the LED peak emission at 635 nm and detector peak response at 670 nm (with 635 nm as the second peak response).

Figure 5 shows the relationship between the coefficient of the determination and individual wavelengths between 830 nm and 930 nm in quantifying the carambola pH and firmness measured using the interactance technique via the Jaz spectrometer. After the measurement using the spectrometer has been completed, similar measurement was then conducted using the OF-RS and OF-NIRS. For the pH measurement, the response algorithm was generated by cubic regression. For the pH measurement, the \( R \) value increased with higher wavelength values, and for the range examined in this paper, the peak was located at 930 nm \((R = 0.784)\). A higher NIR value was closely related to the water absorbance band. For the firmness measurement, the \( R \) value remained relatively the same and fluctuated between 0.783 and 0.789, with the peak located at 890 nm \((R = 0.789)\). The OF-NIRS operated within this wavelength range with the LED peak emission at 880 nm and detector peak response at 900 nm.

Fig. 4 Relationship between the linear correlation coefficient and individual wavelengths between 600 nm and 700 nm in quantifying carambola pH and firmness measured using the interactance technique via the Jaz spectrometer.

Fig. 5 Relationship between the correlation coefficient and individual wavelengths between 830 nm and 930 nm in quantifying carambola pH (quadratic \( R \)) and firmness (cubic \( R \)) measured using the interactance technique via the Jaz spectrometer.

After the measurement using the spectrometer has been completed, similar measurement was then conducted using the OF-RS and OF-NIRS. For the pH measurement, the response algorithm was generated by linear and quadratic regression (Fig.6). The linear representation of the carambola pH
measurement is expressed in (4) with \( R = 0.861 \) and \( RMSE = 0.22 \) pH. The quadratic regression generated better correlation for the pH measurement with \( R = 0.876 \) and \( RMSE = 0.211 \) pH, as expressed in (5).

\[
pH = 0.424 + 0.00162R_{635} \quad (4)
\]
\[
R = 0.861; \ RMSE = 0.220 \text{ pH}
\]

\[
pH = 4.177 – 0.003016R_{635} + 0.000001R_{635}^2 \quad (5)
\]
\[
R = 0.876; \ RMSE = 0.211 \text{ pH}.
\]

4. A higher measurement performance can be expected if the current LED that operates at the peak wavelength of 635 nm is replaced by a higher intensity LED with a peak wavelength at closer to 675 nm (the most suitable wavelength for the measurement of chlorophyll in carambola [41]), which is also proven as the highest efficient wavelength to determine the carambola intrinsic quality. In the research conducted by French et al. [42], they have identified the peak absorbance for chlorophyll at 661.6 nm, 669.6 nm, 677.1 nm, and 683.7 nm. On the other hand, several researchers have concluded absorbance bands for chlorophyll in the determination of the fruits’ intrinsic quality. For instance, wavelengths at 672 nm and 680 nm have been identified by Gomez et al. [18] and Kawano et al. [43] as the peak absorbance for mandarin oranges while 673 nm has been identified by Cao et al. [13] as the peak absorbance for grapes. These stand as the evidence that the developed optical fiber sensor, OF-RS, can also be used to evaluate intrinsic quality parameters of different fruits via their chlorophyll content.

Next, this section will discuss the application of the OF-NIRS which has been applied for the measurement of the carambola pH and firmness. For the measurement of the carambola pH, the response algorithm was best generated only by quadratic regression (Fig. 8) and is expressed in (7) with \( R = 0.751 \) and \( RMSE = 0.293 \) pH.

\[
pH = 5.035 – 0.001221R_{880} + 0.0000001R_{880}^2 \quad (7)
\]
\[
R = 0.751; \ RMSE = 0.293 \text{ pH}.
\]
For the carambola firmness measurement, the response algorithm was best developed through cubic regression (Fig. 9) and is expressed in (8) with $R = 0.773$ and $\text{RMSE} = 1.207 \text{ kgf}$.

$$\text{Firmness} = 18.94 - 0.02705R_{880} + 0.000014R_{880}^2 - 0.0000001R_{880}^3 \quad (8)$$

$R = 0.773; \text{RMSE} = 1.207 \text{ kgf}$.

The OF-NIRS produced satisfactory results for the carambola pH and firmness measurements with $R$ over 0.75. Even if the single system performance results obtained were considered low, the exciting finding from this experiment revealed that the results obtained are very closely related to the measurement produced by individual wavelength analysis using the Jaz spectrometer, as shown in Fig. 5, which were also generated through quadratic and cubic regressions. The combination of the OF-RS and OF-NIRS results for the carambola pH and firmness measurements through multiple linear regression did not produce any improvements in the measurement accuracy. The complete measurement results obtained in this research are listed in Table 3.

### 4. Conclusions

From the comparative analysis, the results obtained through the optical fiber sensors were almost similar to those measured using the Jaz spectrometer. For the measurement of intact carambola pH, the OF-RS showed more accurate results than the Jaz spectrometer at the wavelength of 635 nm. In summary, using spectroscopy, the combination of the optical components of the OF-RS and OF-NIRS has been proven capable of producing optimum results for the measurement of the pH and firmness of carambola, which were comparable with those of the commercial spectroscopy instrumentation. The followings are the advantages of using the optical fiber sensors for the measurement of the fruit intrinsic quality:

1. It has a much lower cost [approximately USD 300 for the optical fiber sensors, OF-RS, and OF-NIRS (without the fiber probe) versus approximately USD 5,000 for the Jaz spectrometer and tungsten halogen lamp (without the fiber probe and sampling facilities)].

2. It can produce the instantaneous reading with only about 2 s of individual measurement execution time. The system can also produce the online reading with execution time of less than 0.1 s. Typical commercial spectrometers require the intensive data processing and application of tedious mathematical analysis of the raw spectra.

3. It is easy to operate with the minimal amount of training. Commercial spectrometers can only be operated by workers well trained in optical instrumentation and software application.

4. It has less complexity in terms of the system structure because it is only operated by LEDs and detectors that contribute directly to the measurement of the fruit intrinsic quality. The application of LEDs

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### Table 3 Summary of results obtained from optical fiber sensors.

| Optical System | Relationship | pH   | Firmness (kgf) |
|----------------|--------------|------|----------------|
| OF-RS (635 nm) | Linear       | $R = 0.861$ | $R = 0.872$ |
|                |              | $\text{RMSE} = 0.220$ | $\text{RMSE} = 0.909$ |
|                | Quadratic    | $R = 0.875$ | $\text{RMSE} = 0.211$ |
| OF-NIRS (880 nm) | Quadratic | $R = 0.751$ | $\text{RMSE} = 0.293$ |
|                | Cubic        | $R = 0.773$ | $\text{RMSE} = 1.207$ |
| Jaz spectrometer (635 nm) | Linear   | $R = 0.815$ | $R = 0.0873$ |
|                |              | $\text{RMSE} = 0.251$ | $\text{RMSE} = 0.907$ |
|                | Quadratic    | $R = 0.819$ | $\text{RMSE} = 0.251$ |
| Jaz spectrometer (880 nm) | Quadratic | $R = 0.767$ | $\text{RMSE} = 0.330$ |
|                | Cubic        | $R = 0.784$ | $\text{RMSE} = 1.271$ |
has eliminated the need for bandpass filters. Commercial spectrometers have a broad array of wavelengths that are neglected because they do not contribute to the developed measurement algorithm.

Further research can be expanded in several areas to increase the efficiency of the optical fiber sensor. The application of LEDs with various characteristics such as different wavelengths within the visible and NIR regions and narrower spectral width and higher emission intensity, especially within the NIR region, could identify the best configuration that can generate higher measurement accuracy. The optical fiber system with sequential LED triggering from one wavelength to another will be required if the measurement of multiple wavelengths is made at precisely the same location on the intact fruit sample. The success of the specialized optical fiber sensor for the measurement of the fruit intrinsic quality can positively benefit fruit growers and allow the transformation of the current definition of the fruit quality from the physical attribute-based evaluation to intrinsic assessment.

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