Tomatoes Maturation Analysis with Reflectance Spectral Images

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Abstract: The demands for high quality products are increasing in the market, so several tests are performed. These analyses are destructive, time consuming and sometimes expensive. The spectral imaging technique is an alternative that can reduce costs and time and increase the efficiency and cost-effectiveness of these analyses. In this context, the aim of this study was to verify the spectral behavior of the tomato in function of the maturation and its potential use for non-destructive evaluation of food quality. Eighty tomatoes of the Italian group were selected in four stages of maturation (green, pink, light red and red). Images were collected in wavelengths 480 nm to 710 nm (every 10 nm). After the images were collected, the fruits were submitted to the physical-chemical analyses. The following parameters were analyzed: color, color index (CI), pH, soluble solids (SS), total titratable acidity (TTA) and ratio (SS/TTA). After characterization the mean spectrum was obtained by the extraction of the spectral characteristics. It was observed that the spectral behavior was the same for all stages of maturation. It was possible to differentiate the green tomatoes in the wavelengths from 520 nm to 560 nm and from 610 nm to 660 nm. These intervals represent the area of reflection of chlorophyll and lycopene, respectively. Near the wavelength 560 nm, the highest absorbance was recorded by ripe tomatoes due to the concentration of lycopene. It is concluded that the best wavelengths for analysis are those in which green has the highest reflectance and it is possible to apply this technique for non-destructive analysis of the maturation of tomatoes.

Key words: Non-destructive analysis, mean spectrum, physical-chemical analysis, Solanum lycopersicum.

1. Introduction

Currently, tomato is considered one of the most important vegetables in the world. In 2012, China was the largest tomato producer and Brazil ranked eighth in the world [1]. The tomato for processing depends on several characteristics and good quality. Laboratory tests are performed to ensure nutritional and health standards. These tests are destructive, time consuming, and sometimes expensive chemical reagents and equipment are required, which makes the procedure very costly. As a result, studies are being conducted to look for faster, more accurate and more cost-effective data acquisition methods and tools.

Recent technological advances have increased the potential of light-tissue interaction in studies of photosynthetic activity and chlorophyll content in plants [2], monitoring of product quality and safety [3, 4] and detection of plant injuries [5, 6], maturity stages [7], physical characteristics and chemical components, such as sugar content [8-10].

One such technique, spectral imaging, can be used as an alternative to improve control of tomato processing, reduce costs and ensure quality products to the consumer. This technique generates spatial maps that allow the identification and distribution of several sample components and can be applied for various purposes [11]. However, for tomatoes, it is necessary to select and define the best spectral bands and the most appropriate way to develop the image
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acquisition and processing system.

In this context, the objective of this work was to verify the spectral behavior of the tomato in function of the degree of maturation and its potential use for non-destructive evaluation of food quality.

2. Materials and Methods

2.1 Samples and Physical-Chemical Analysis

The samples were obtained from a fruit and vegetable distribution center in the city of Montes Claros—Minas Gerais, Brazil. Eighty tomatoes of the Italian group were selected in four stages of maturation (green, pink, light red and red).

The physical-chemical characteristics analyzed were color, color index (CI) [12], pH, soluble solids (SS), total titratable acidity (TTA) [13] and ratio (SS/TTA).

Only SS data presented normal distribution and were analyzed by the Tukey test at 1% significance. The other characteristics, which presented abnormal distribution, were submitted to the Kruskal-Wallis test at 1% significance level. The software used for statistical analysis was R.

2.2 System and Acquisition of Images

The image acquisition system was composed of monochromatic charged couple camera (CCD), model mvBlueFOX-223G Matrix Vision, Germany, filter wheel (model FW102B, Thorlabs, USA), 25 mm objective lens and light emitting diodes high-power white (LEDs) to promote the excitation source of the spectral images. In addition, for image acquisition and lighting control, an Arduino® Uno was used.

The images were captured using wavelengths between 480 nm and 710 nm (models from FB480-10 to FB710-10, Thorlabs, USA), every 10 nm to obtain spectral response of the product at different wavelengths. Therefore, 24 images were captured for each sample, totaling 1,920 images.

2.3 Data Processing

After collecting the images, to facilitate the following analyses, they were processed through operations aimed at improving the visibility of the characteristics of interest and correcting defects of light scattering and possible noise [14, 15].

The images were first normalized according to the camera and optical filter transmission curves. After normalization, the regions of interest (ROIs) were highlighted manually.

A dataset composed of a matrix was created, in which the lines present the pixel values and the columns the wavelength for each stage of maturation of the tomato. Once the dataset was done, the spectral data were preprocessed.

In order to reduce environmental and texture effects, known as additive and multiplicative effects, and to smooth noise, the data were preprocessed using a relative normalization. The relative normalization consists of calculating the sum of the spectrum and then subtracting from the mean and dividing by the sum of each value of the spectrum. Thus, the spectral scattering effect was removed and, therefore, a normalized dataset was obtained [16].

From the normalized dataset, the mean spectrum of each maturation stage of the tomato samples was obtained from the mean and standard deviation of the wavelengths.

The image analysis was performed with the help of MATLAB® software.

3. Results and Discussion

3.1 Chemical Characteristics at Different Maturation Stages

It was observed that SS and SS/TTA values are in high quality fruit standards as stipulated by Mencarelli and Saltveit Jr. [17]. However, tomatoes classified as light red and red had lower TTA values than the ideal one (Table 1). These characteristics directly interfere
with the taste of the product and the use of additives in processing, so they should be analyzed together. The tomatoes showed pH around 4.0, which is common for this fruit [18-20].

### 3.2 Color Parameters at Different Maturation Stages

The only color parameter that was not influenced by maturation was $b^*$, which represents the color variation from yellow to blue (Table 2). In relation to the luminosity ($L^*$), green tomatoes presented the highest value and there was no difference between light red and red tomatoes. This is due to the concentration of chlorophyll and lycopene in the green and mature fruits, respectively. There is a negative correlation between lycopene and this color parameter, so with the advancement of maturation the tendency of $L^*$ is decreased [21, 22]. During the ripening of the tomato, there is intense degradation of chlorophyll and lycopene synthesis [23]. When the compounds being responsible for the red color, mainly lycopene, begin to be synthesized, the value of $L^*$ decreases, indicating darkening of the fruit [24].

The behavior observed for the variable $a^*$ (variation from green to red) was the same as for the CI variable. As the tomatoes mature, the values of $a^*$ and CI increased, however the values for light red and red tomatoes are the same. During tomato maturation, the variation from green to red ($a^*$) is intense and ends up masking the other parameters of the CI [24]. Lycopene is also related to $a^*$, so the increase in the values of $a^*$ indicates higher lycopene contents [21].

### 3.3 Spectral Behavior at Different Maturation Stages

The difficulty of differentiating light red and red tomatoes was also observed in the reflectance mean spectrum (Fig. 1), which confirmed the relationship between reflectance and physical-chemical characteristics. From 520 nm to 560 nm and from 610 nm to 660 nm, only green tomatoes can be distinguished. This behavior is related to the spectral areas of reflection and absorbance of chlorophyll, respectively.

Liu et al. [25] also observed a region of low reflectance near 660 nm due to the absorption of chlorophyll. Clément et al. [26] observed a peak of reflectance for green tomatoes between 550 nm and 560 nm, because in this region there is no absorption by chlorophyll. Near 680 nm was observed that the absorption was related to the green coloration of the fruits, which was influenced by the chlorophyll content [26, 27].

It is noted that at wavelengths 560, 580, 630, 640,
650 nm, the reflectance spectra follow an order according to maturation. Liu et al. [25] also obtained a graph in which it was possible to observe this arrangement in some wavelengths. This can be explained by the variation in the concentration of chlorophyll and lycopene throughout the maturation, which influences the reflectance of the tomato.

The spectral behavior was similar for the maturation stages, with the exception of the interval between 600 nm and 620 nm, where the reflectance of green tomatoes decreases and that of the others increases. The reflectance peaks at wavelengths 560 nm and 580 nm, spectral regions of green, occurring due to chlorophyll reflectance and lycopene absorbance, compounds present in higher concentrations in green and mature fruits, respectively. At 620 nm of spectral region of red, there is a reflectance of lycopene and chlorophyll absorbance [28], which explains the higher reflectance obtained for fruits with more advanced maturation.

Among the wavelengths 510 nm and 580 nm of chlorophyll reflection region [26, 29], the highest reflectance value is presented by green tomatoes and the lowest by light red and red tomatoes. Between 610 nm and 660 nm, the behavior is the opposite; the highest values are observed for tomatoes with advanced maturation stage and the lowest for green tomatoes.
At the wavelength 565 nm, the highest absorbance was recorded by mature tomatoes (Fig. 2), as it is in this region that there is high absorbance of lycopene in aqueous medium [28].

The images of reflectance were efficient to differentiate green tomatoes from the others, but it was not possible to distinguish graphically tomatoes with more advanced maturation, which can be possible through multivariate analysis and using other sources of illumination. It is verified that the spectral response of the fruit depends on its chemical characteristics, mainly chlorophyll and lycopene content, which give green and red color to the fruits, respectively. These characteristics are evidenced in the reflection and absorption peaks of chlorophyll and lycopene. Therefore, the technique is applicable to determine the maturation of the fruits, but other techniques of analysis must be combined to apply spectral images in the evaluation of food quality.

4. Conclusions

The spectral behavior of tomato is related to the maturation of the fruit. Green tomatoes showed higher reflectance between wavelengths 520 nm and 560 nm. On the other hand, mature tomatoes presented higher reflectance between the wavelengths 610 nm and 660 nm. This knowledge is very important to improve the control of tomato processing, reduce costs and ensure quality products in a non-destructive way. The tomatoes follow a spectral pattern independent of the maturation stage, and the best wavelengths to distinguish green tomatoes are from 520 nm to 560 nm and from 610 nm to 660 nm. However, it may be that multivariate analysis techniques can determine the other stages of maturation. Therefore, it is possible to use spectral images for non-destructive evaluation of tomato maturation, but for the method to be applied for analysis of food quality, other techniques must be used together.

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