Image analysis of oxidation induced food colour changes

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Abstract

Colour evaluation, by means of multivariate image analysis was applied on colour lightness changes corresponding to oxidative browning of minimally-processed basil-based food emulsions. We have analysed the influence of metal-chelating protein lactoferrin on browning of emulsions (attributed to enzymaic and non-enzymatic oxidation). The colourgram of digitalized images, taken during oxidation of the samples, was evaluated, both by colour analysis and by PCA analysis. The basic colour information was derived from frequency colour distribution for RGB and HSV colour system, while the PCA analysis was carried out by evaluating covariance matrix, for raw, mean centered and autoscaled matrix of the basic R, G, B colour information, and the evaluation of eigenvalues and eigenvectors for all three numeric models. The resulting colourgram information is one-dimensional signal, which consists of 4900 elements, and describes colour properties of the image. The changes in material due to oxidation was observed by the changes in lightness frequency distribution (L).

Key words: colourgram, PCA analysis, food, oxidation, browning

1. Introduction

In the assessment of food colour it is common practice to use spectrophotometers or light sensitive cells to quantify colour characteristics. The most diffused instruments evaluate only restricted areas of food samples or overall light reflectance from the entire surface of the food matrix, thus being not appropriate for inhomogenous food products. The changes in colour, occuring due to oxidation processes of such products can be observed by evaluation algorithm describing the colour intensity of digitalised images (Antonelli, A., et al. 2004, Riva M., et al. 2005, Acosta B., et al. 2010, Quevedo R., et al. 2009a, Kumara B., 2003, Gazmuri A.M., et al. 2009, Pilin A., et al. 2009, Tien C. L., et al. 2001, Quevedo R., et al. 2009b). These authors evaluated colour characteristics of various materials. The colour images were taken using colorimeters, CCD cameras or optical scanners. Authors used different colour coordinate systems, and colour features were evaluated by image analysis.

Basil-based spread, similar to non heat-processed “pesto”, an inhomogenous Italian pasta sauce, whose main ingredients are basil, olive oil, cheese and walnuts, is identified as an ideal benchmark to test the antioxidative properties of lactoferrin, metal-chelating protein, isolated from milk, which is an antimicrobial agent of broad spectrum, and, also, influence lipid
oxidation (Chiu, C.H., et al. 2007, Huang, et al. 1999, Wakabayashi H., et al. 2006, Naidu A. S., 2002, Al-Nabulsi A. A., et al. 2005, Al-Nabulsi A. A., et al. 2007, C. Jacobsen, et al. 2008). Addition of natural antioxidative agents potentially influences lipid oxidation and browning (attributed to combined enzymatic and non-enzymatic oxidation) of minimally-processed green vegetable-based food emulsions. Pesto, obtained by the traditional recipe, from fresh basil, is limited to the very short shelf life due to oxidative processes and microbial load (Severini, C., 2008, Fabiano, B., et al. 2000). Oxidative browning of pesto samples was determined by colour lightness changes, applying colour evaluation, by means of multivariate image analysis. Software was developed for processing of digitalized images in the form of one-dimensional signal, describing the colour content. The coloured images were taken with a common digital CCD camera. The basic colour information was derived from frequency colour distribution for red, green and blue (RGB) and hue, saturation and intensity (HSI) colour system, while the PCA analysis (Huang, J., et al. 2003, Shlens, 2009) was carried out by evaluating covariance matrix, for raw, mean centered and autoscaled matrix of the basic R, G, B colour information, and the evaluation of eigenvalues and eigenvectors for all three numeric models. The changes in material due to oxidation was observed by the changes in lightness frequency distribution (L).

2. Materials and methods

2.1 Preparation of basil-based spreads

Basil-based spreads similar to Italian pesto sauce, were prepared from fresh basil which was treated with hydrogen peroxide and/or short temperature shock during washing. The product contained virgin olive oil, refined sunflower oil, cheeses, such as parmesan and kachkaval, whey, walnuts, sunflower seeds, salt and garlic, as well as organic acids (ascorbic, citric and lactic acids) (Sigma-Aldrich, Germany) and lactoferrin (bovine lactoferrin, 95% protein, DMV International, Veghel, The Netherlands) as regulators of acidity, antioxidative and antimicrobial agents. The product was packaged in glass jars of 150g.

Nine different spreads were prepared for color change evaluation experiments, as shown in Table 1. These spreads were obtained by modulating the concentration of ascorbic acid, citric acid, lactic acid, NaCl and lactoferrin.
Table 1. Treatments of basil-based “pesto” spreads. Nine spreads were obtained by modulating levels of ascorbic acid, citric acid, lactic acid, NaCl and lactoferrin.

2.2 Colour analysis

Colour images of basil-based pesto spreads were captured by a Sony PowerShot A550 and A590 CCD camera, which is a common digital camera for home use. All the acquired images were 24 bit RGB (16.8 millions of colours) with a 1024 x 768 spatial resolution. The macro function of the digital camera has been used, to cover a scene area of approximately Ø10 cm. Each pesto sample was smeared to an approximately constant 3-5 mm thickness and to an area sufficient to cover completely the image scene, in order to avoid the presence of background in the image. Then, it was placed on a white paper napkin set on a flat white painted surface, 15cm below the digital camera. Paper napkins were used in order to absorb the excess of oil and therefore to avoid undesired reflection effects. With this setup, it was possible to capture images with negligible shadows and without specular reflections. There were 9 different pesto formulations to be observed, and the images of each different formulation of pesto were taken periodically by camera 22 times during the period of approximately 320 minutes.

The acquired RGB images were transferred to a personal computer in the form of jpeg compressed image files by means of the software provided together with the digital camera. The size of the imported jpeg images ranged from 900 to 1250 KB for the 9 x 22 = 198 images of the data set. For the elaboration of the images, every image file is imported in the originally developed computer program as a scanned RGB picture (this program is developed using Microsoft Visual Basic 6 software). This program transforms digital picture data to the three-dimensional array, using simple Visual Basic procedure:
C = Picture1.Point(X, Y): ‘3x8 bit color point at X, Y coordinate
Blue = Int(C / 65536): C = C - Blue * 65536
Green = Int(C / 256)
Red = C - Green * 256

The first step in the elaboration of every single digital RGB image with size \{r, c, R, G, B\} (where r is the number of pixel row, c is the number of pixel column and R, G and B channels) consists of unfolding it to a \{(r x c), (R, G, B)\} two-dimensional matrix of the R, G and B values (variables) for the r x c pixels (objects), i.e., \(r \times c\) x 3 dimension. Since the imported files are 24 bit colour images, each one of the three R, G and B "slices" of the 3D array are \(24/3\) 8 bit greyscale images, and therefore the three R, G and B variables can assume all the integer values in the range 0-255. After this first step these coordinates were normalized, i.e., divided by 255, in order to be represented as values varies from 0 – 1.

Then, a series of quantities derived directly by the normalized R, G and B variables with simple algebraic operations are calculated for every pixel: lightness (L), which is the sum of R, G and B, divided by 3 \(L = (R+G+B)/3\) and the three relative colours, i.e. relative red \(RR = R/(L \times 3)\), relative green \(RG = G/(L \times 3)\) and relative blue \(RB = B/(L \times 3)\), which are ratios between the R, G, B values and their sum, L. It should be noticed that these values are also normalized, i.e., varies from 0 to 1. The relative values have been considered since a particular colour, e.g. red colour does not correspond to a high absolute value of the corresponding R variable, but to a high value of the ratio between R and L. For example, we can have R = 255 for pure red (with G = 0 and B = 0), where RR coordinate equals 1.000, for white (with G = 255 and B = 255), RR coordinate reaches 0.333, and for yellow (with G = 255 and B = 0), RR coordinate is 0.667, etc.

Corresponding program routine should be as follows:

\[
\begin{align*}
L &= B + G + R \\
RR &= R / L; RG = G / L; RB = B / L: \text{‘relative coordinates} \\
L &= L / 3: \text{‘lightness}
\end{align*}
\]

The hue, saturation and intensity (HSI) values are also calculated, using the developed computer program, for the conversion from the RGB to the HSI colour space. The I values are computed for every pixel as the maximum value between R, G, and B:

\[
I = \max(R, G, B)
\]  

while the S values are given by:

\[
S = \frac{\max(R, G, B) - \min(R, G, B)}{\max(R, G, B)}
\]

\(H\) value is defined by a quite complex algorithm, difficult to represent in an easily readable analytical form, in a way that it varies from 0 to 1 with the corresponding colour passing from red through yellow, green, cyan, blue, and magenta, and then back to red. The Visual Basic procedure should be expressed as follows:

\[
\begin{align*}
x1 &= (G - R) / \text{Sqr}(2); y1 = (2 * B - R - G) / \text{Sqr}(6) \\
\text{If } x1 = 0 \text{ And } y1 = 0 \text{ Then } Arg = 0 \\
\text{If } x1 = 0 \text{ And } y1 >= 0 \text{ Then } Arg = \pi / 2
\end{align*}
\]
If \( y_1 = 0 \) And \( x_1 < 0 \) Then Arg = \( \pi \)
If \( x_1 = 0 \) And \( y_1 < 0 \) Then Arg = \( -\pi / 2 \)
If \( x_1 > 0 \) Then Arg = \( \arctan(y_1 / x_1) \)
If \( x_1 < 0 \) And \( y_1 \geq 0 \) Then Arg = \( \pi - \arctan(-y_1 / x_1) \)
If \( x_1 < 0 \) And \( y_1 < 0 \) Then Arg = \( -\pi + \arctan(-y_1 / -x_1) \)
\( H = \text{Arg} \times 180 / \pi + 150 \)
If \( S = 0 \) Or \( I = 0 \) Then \( H = 0 \)
If \( H < 0 \) Then \( H = H + 360 \)
If \( H \geq 360 \) Then \( H = H - 360 \)

After normalization of \( H \) value (\( \frac{H - H_{\text{min}}}{H_{\text{max}} - H_{\text{min}}} \)), all above mentioned variables \((R, G, B, L, RR, RG, RB, H, S, I)\) fall within the range of 0 to 1 (all of them are normalized, as mentioned above). Each of this value can be represented by discrete number: \( j = \frac{i}{255} \), where, \( i = 0 \) to 255. Then, a frequency distribution vector with a length of 256 elements is calculated over all the single pixels \((r \times c)\), for each one of the considered variables: \( R, G, B, L, RR, RG, RB, H, S, I\), for instance the elements of red value distribution curve are calculated as follows:

If \( R = j \) then \( R_1(j \times 255) = R_1(j \times 255) + 1 \)

The elements of all other variables are calculated similarly. There were 10 variables, i.e., 10 distribution curves with 256 elements for each picture \((256 \times 10 \text{ matrix})\).

The PCA analysis was evaluated in above mentioned program, using the score matrix of raw unfolded matrix, the score matrix of mean centered score unfolded matrix, and the score matrix of autoscaled unfolded RGB matrix. The element of raw matrix were: \( R(i), G(i), B(i), \) \( i = 0, \ldots, 255 \). Mean centered matrix was evaluated by evaluating the subtracting the mean value of each color vector to each element of the vector:

For \( i = 0 \) to 255: \( R_{\text{MEAN}}(i) = R_1(i) - R_{\text{MEAN}} \)

The autoscaled matrix was evaluated by dividing the mean centered matrix vectors with standard deviations for each vector:

For \( i = 0 \) to 255: \( R_{\text{AUTOSC}}(i) = R_{\text{MEAN}}(i) / R_{\text{STD}} \)

The evaluation of covariance and correlation matrix for raw, mean centered and autoscaled matrix followed, with the evaluation of eigenvectors and eigenvalues of covariance matrix. The eigenvectors of covariance matrix for all three models were treated as loading vectors in PCA analysis. The used equations for calculation of covariance and correlation were:

\[
\text{cov}_{RG}(R,G) = \frac{\sum_{i=0}^{255} (R(i) - R_{\text{MEAN}})(G(i) - G_{\text{MEAN}})}{255}
\] (3)
\[ corr_{RG} (R,G) = \frac{\sum_{i=0}^{255} (R(i) - \text{RMEAN})(G(i) - \text{GMEAN})}{\sqrt{\sum_{i=0}^{255} (R(i) - \text{RMEAN})^2 \cdot (G(i) - \text{GMEAN})^2}} \] (4)

The eigenvalues (\( \lambda \)) of evaluated covariance matrix were calculated, as usually, using the equation:

\[
COV = \begin{bmatrix}
1 & \text{cov}_{RG} & \text{cov}_{RB} \\
\text{cov}_{RG} & 1 & \text{cov}_{GB} \\
\text{cov}_{RB} & \text{cov}_{GB} & 1
\end{bmatrix}, \quad \det(COV - \lambda \cdot I) = 0
\] (5)

where, \( COV \) is the covariance matrix, and \( I \) is identity matrix.

The evaluation of roots for equation (5), was implemented in above mentioned program, using Newton’s iterative method for finding roots of function 

\[-a\lambda^3 + b\lambda^2 + c\lambda + d = 0 :\]

\[
x = ((3*c/a) - (b^2/2a^2))/3; \quad y = ((2*b^3/a^3) - (9*b*c/a^2) + (27*d/a))/27
\]

\[
z = y^2/4 + x^3/27
\]

\[
i = \text{sqr}(y^2/4 - z); \quad j = -i^(1/3); \quad k = \text{arccos}(-y/2*i)
\]

\[
m = \text{cos}(k/3); \quad n = \text{sqr}(3)\text{*sin}(k/3); \quad p = -(b/3*a)
\]

\[
\text{Eig1} = -2j*m + p; \quad \text{Eig2} = j*(m + n) + p; \quad \text{Eig3} = j*(m - n) + p
\]

The eigenvectors of covariance matrix are then calculated as:

\[ P = COV - \lambda \cdot I \] (6)

and it should be calculated for all three models, raw, mean centered and autoscaled.

The score matrix is evaluated as:

\[ T = X \cdot P \] (7)

where \( X \) is the raw, mean centered and autoscaled data matrix, \( T \) is the score model and \( P \) is the loading vectors (eigenvalues) of raw, mean centered and autoscaled model.

Equation (7) is used for matrix transformation of original matrix \( X \) (for all three models), and this transformation is defined by \( P \) matrix. The result of this matrix transformation is designed as score matrix for all three models, marked as \( Ps1, Ps2, Ps3, Pm1, Pm2, Pm3, Pa1, Pa2 \) and \( Pa3 \) on colourgram. Each of these matrix is formatted by \( 3 \times 256 \) values (\( R, G, B, \) for raw matrix, mean centered matrix, and autoscaled matrix). These vectors are joined in \( 256 \times 9 \) matrix, and added to above formed \( 256 \times 10 \) matrix.

**Results**

The final form of colourgram is done after normalisation of loading vectors for all three models (each of this matrix is defined by \( 3 \times 3 \) values), and eigenvalues of three PCA models (each model has three eigenvalues).
The resulting colourgram information is one dimensional signal, which is consisted of \(10 \times 256 + 9 \times 256 + 3 \times 3 \times 3 + 3 \times 3 \times 1 = 4900\) elements, as shown above, and it describes the colour properties of the image. The typical form of colourgram is shown on Fig. 1.

\[
\begin{align*}
\text{Fig. 1. Typical colorogram of pesto sample.}
\end{align*}
\]

In this work, only lightness \((L)\) was used to represent data recorded during the experiment, and these data is represented in relative form, i.e., lightness was divided by initial lightness:

\[
L_{\text{rel}} = \frac{L}{L_{\text{init}}}
\]  

(8)

The value \(L\) in equation (8), represents the maximum value reached in frequency distribution vector of lightness (while colourgram evaluation). It is been found that the pesto sample is getting darker while exposed to oxygen (browning effect), according to exponential law \((y = y_0 + A \cdot \exp(x/t))\). Evaluated correlation coefficient for these curves were \(R^2 = 0.87\), and square \(\chi^2 = 1.6\).
**Fig. 2.** Relative lightness of basil-based (*pesto*) spreads during oxidation at room temperature. The effect of 9 treatments with ascorbic, citric, lactic acid, NaCl and lactoferrin, as presented in Table 1.

*Pesto*, obtained by the traditional recipe, from fresh basil, is limited to the very short shelf life. For this reason, pasteurisation or sterilisation and/or lowering of pH and addition of antimicrobial and antioxidative agents are needed for industrial production, but these treatments lead to colour, taste and flavour degradation. Changes in colour of *pesto*, occurring due to oxidation processes (oxidation of phenols, degradation of chlorophyll and non-enzymatic browning, due to the reaction of oxidated lipids with proteins), can be observed by evaluation algorithm describing the colour changes during oxidation process, obtained through colourgram analysis. The relative colour lightness, Figure 2., in all samples with 0.5 g/kg of ascorbic acid containing lactoferrin concentrations 80, 400 and 2000 mg/kg (treatments 2 and 3 and 4, respectively), decreased more quickly during the time, i.e. the browning was more intensive than in the control treatment 1. Samples containing lactic acid (3.6%) in combination with ascorbic acid (1 g/kg) and citric acid (1.5 g/kg), browned abruptly (treatments 8 and 9), after a period of 150 min, while the samples containing ascorbic acid (1 g/kg) and citric acid (1.5 g/kg), (treatments 6 and 7), proved to be least prone to oxidation. Treatments 7 and 9, contained 400 mg/kg of lactoferrin. The comparable results are obtained in different experimental systems, such as the application of digital image processing, which was very useful for studying of enzymatic and non-enzymatic browning of fruits (Riva M., et al. 2005), and L coordinate of digital image was found to be significant for enzymatic browning (Acosta B., et al. 2010).

**Conclusion**

Colour analysis of minimally-processed fresh basil-based food emulsions, by means of multivariate image analysis, was applied on colour lightness changes corresponding to oxidative browning of the product treated with an antioxidative agent, metal chelating protein, lactoferrin.
The colourgram was evaluated, both by colour analyzing and by PCA analysis. The basic colour information was derived from frequency colour distribution for RGB and HSV colour system, while the PCA analysis was carried out by evaluating covariance matrix, for raw, mean centered and autoscaled matrix of the basic R, G, B colour information, and the evaluation of eigenvalues and eigenvectors for all three numeric models. The resulting colourgram information is one-dimensional signal, which is consisted of 4900 elements, and describes colour properties of the image. The changes in material due to oxidation was observed by the changes in lightness frequency distribution (L).

Ascorbic and combination of ascorbic and citric acids inhibited browning of the product, (attributed to combined enzymatic and non-enzymatic oxidation), while lactoferrin slightly enhances browning in the presence of ascorbic acid (0.5 g/kg) at pH 5.2-5.3.

Acknowledgments - The authors are grateful to Ministry of Science and Technological Development of Serbia for the support (projects TR31055 and TR31093).
ИЗВОД

АНАЛИЗЕ СЛИКА ПРОМЕНЕ БОЈЕ ХРАНЕ УСЛЕД ОКСИДАЦИЈЕ

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РЕЗИМЕ

Оцењивање боје, преко вишепараметарске анализе слика примењено је за праћење промена у осветљености боја, које одговарају оксидативном тамњењу минимално-праћених прехрамбених емулзија на бази босилка. Анализирали смо утицај лактоферина, протеина хелатора метала на тамњење емулзија (које може потичи од ензиматског и неензиматског тамњења). Оцењивани су колорограми дигитализованих слика, добијених током оксидације узорака, коришћењем анализе боје и ПЦА анализом. Основна информација о боји добијена је из расподеле учесталости боја, за РГБ (црвено, зелено и плаво) и ХСВ (сјај, засићеност и интензитет) системе боја, док је ПЦА анализи извршена вредновањем коваријансне матрице за основне, осредњене и аутоскалиране матрице почетних РГБ информација о боји и вредновањем сопствене вредности коваријансне матрице и вектора сопствене вредности коваријансне матрице за сва три нумеричка модела. Резултирајућа информација колорограма је једномернцијала сигнал који се састоји од 4900 елемента и описује особине боје слике. Промене у материјалу, које настају током оксидације праћене су преко промена у расподели учесталости осветљености (Л).

КЛЮЧНЕ РЕЧИ: колорограм, ПЦА анализи, храна, оксидација, тамњење

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