# Applying LabRGB to Real Multi-Spectral Images 

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#### Abstract

A spectral encoding/decoding method, called LabRGB, was proposed last year [1]. LabRGB consists of six unique base functions and physically meaningful encoding values. LabRGB has several features; color characteristics can be estimated by its encoding values, fixed base functions need not exchange beforehand and tri-chromatic images as well as multi-spectral images can be handled by a single form. In this paper, the derivation of LabRGB was described in detail including fast straightforward calculation of encoding/decoding calculation. Then LabRGB was applied to real multi-spectral images. LabRGB performed well in encoding/decoding various types of multi-spectral images and confirmed to be suitable for practical use.


## Introduction

Spectral encoding/decoding using eigenvectors is a well-known method since a long time ago. Generally speaking, it shows the least estimation error. On the other hand, eigenvectors cannot be defined uniquely, because they depend on a sample selection of population. Also, encoding/decoding values using eigenvectors have no physical meaning. So it is difficult to directly estimate either a shape of spectral reflectance, or color characteristics of an original object color. It is therefore not easy to verify an encoding / decoding process. Furthermore, it cannot be applied to current tri-chromatic imaging systems directly. As such, several challenges have been made on an encoding/decoding method described above. A recently reported one is $\operatorname{LabPQR}$ [2]. $L a b P Q R$ is a concept of the encoding which has three dimensions (CIELAB [3]) to represent the colorimetric characteristics of a color under a specific illuminant and additional dimensions ( $P Q R$ ) to describe the metameric black spectrum of a spectral power distribution [4]. The intention of $L a b P Q R$ is to convey physical values so that an encoding value can be used to estimate an original object color. Several variations of the $P Q R$ aspects of $\operatorname{LabPQR}$ have been described in the literature [2], [5] including those based on a population of samples or those based on fundamental spectral stimuli4. Aforementioned $L a b R G B$ is another variation of the $L a b P Q R$ concept. $L a b R G B$ has unique, well-defined base functions, physically meaningful encoding values, and are cable of handling both spectral imaging and current tri-chromatic imaging equipments. LabRGB encoding/decoding process has been described in the previous paper [1].
$L a b R G B$ encoding/decoding is done with the following way. The base functions set chosen as shown in Eq. 1. The first three base functions are roughly designed to represent $R G B$ spectral distribution curve and the last three base functions cover higher frequency.

$$
\left.\begin{array}{l}
e_{1}(\lambda)=\sin \left(\frac{1}{2} \pi \frac{\lambda-\lambda_{\min }}{\lambda_{\max }-\lambda_{\min }}\right) \\
e_{2}(\lambda)=\cos \left(\frac{1}{2} \pi \frac{\lambda-\lambda_{\text {min }}}{\lambda_{\max }-\lambda_{\text {min }}}\right) \\
e_{3}(\lambda)=\sin \left(\pi \frac{\lambda-\lambda_{\text {min }}}{\lambda_{\text {max }}-\lambda_{\text {min }}}\right)  \tag{1}\\
e_{4}(\lambda)=\cos \left(\frac{3}{2} \pi \frac{\lambda-\lambda_{\text {min }}}{\lambda_{\text {max }}-\lambda_{\text {min }}}\right) \\
e_{5}(\lambda)=\sin \left(2 \pi \frac{\lambda-\lambda_{\text {min }}}{\lambda_{\text {max }}-\lambda_{\text {min }}}\right) \\
e_{6}(\lambda)=\cos \left(\frac{5}{2} \pi \frac{\lambda}{\lambda_{\text {max }}-\lambda_{\text {min }}}\right)
\end{array}\right\}
$$

With this base functions set, color characteristics can be estimated by weighting factors of the six base functions. The shape of the trigonometric base functions are shown in Fig. 1.

Spectral reflectance estimation was made using an equation obtained by substituting Eq. 1 into the following Eq. 2.

$$
\begin{equation*}
\tilde{\rho}(\lambda)=\sum_{i=1}^{6} w_{i} \cdot e_{i}(\lambda) \tag{2}
\end{equation*}
$$

Where, $\lambda$ is wavelength, $\hat{\rho}(\lambda)$ is spectral reflectance estimation of an object color, $e_{i}(\lambda)$ is $i$-th base function and, $w_{i}$ is a weighting factor of the $i$-th base function.


Figure 1. LabRGB base function
$L a b R G B$ uses a combination of $C I E L A B$ and $R G B$. Encoding is done by the following steps.
a) Calculate $C I E X Y Z$ and $C I E L A B$ [6] values of spectral reflectance of an object color $\rho(\lambda)$ using CIE1931 2 degree observer
b) Substitute Eq. 1 and $\rho(\lambda)$ into Eq. 2 and calculate 6 weighting factors $w_{1} \sim w_{6}$ for 6 base functions by multiple regression analysis
The actual LabRGB encoding values $\operatorname{are} L^{*} a^{*} b^{*} w_{1} w_{2} w_{3}$.

Decoding is done by the following steps.
a) Calculate an estimation of CIEXYZ values $\hat{X} \hat{Y} \hat{Z}$ using only $w_{1} w_{2} w_{3}$ as,

$$
\left.\begin{array}{l}
\hat{X}=\sum_{i=1}^{3} w_{i} \int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{x}(\lambda) d \lambda  \tag{3}\\
\hat{Y}=\sum_{i=1}^{3} w_{i} \int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{y}(\lambda) d \lambda \\
\hat{Z}=\sum_{i=1}^{3} w_{i} \int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{z}(\lambda) d \lambda
\end{array}\right\}
$$

b) Calculate $X Y Z$ values from the first three LabRGB encoding values $L^{*} a^{*} b^{*}$, then calculate $w_{4} w_{5} w_{6}$ from $X Y Z$ values and estimated $\hat{X} \hat{Y} \hat{Z}$ values using Eq. 4 by solving three dimensional $1^{\text {st }}$ order equation

$$
\left.\begin{array}{l}
X-\hat{X}=\sum_{i=4}^{6} w_{i} \int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{x}(\lambda) d \lambda  \tag{4}\\
Y-\hat{Y}=\sum_{i=4}^{6} w_{i} \int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{y}(\lambda) d \lambda \\
Z-\hat{Z}=\sum_{i=4}^{6} w_{i} \int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{z}(\lambda) d \lambda
\end{array}\right\}
$$

c) Substituting obtained $w_{4} w_{5} w_{6}$ and the three encoding values $w_{1} w_{2} w_{3}$ into Eq. 2 to calculate spectral reflectance estimation of an object color $\hat{\rho}(\lambda)$
LabRGB showed almost equal performance to traditional orthogonal eigenvector method in spectral estimation, and even better performance in colorimetric estimation.

The current LabRGB involves multiple regression calculation in encoding process and three dimensional $1^{\text {st }}$ order equations in its decoding process. These two calculations involved fair amount of computation time.

This paper describes the derivation of fast straightforward $L a b R G B$ calculation and the results of applying $L a b R G B$ to real multi-spectral images.

## Straightforward encoding by multiple regression coefficients

Residual error of a spectral reflectance curve $D$ can be written as Eq. 5, where, $\rho(\lambda)$ is spectral reflectance of an object color and $\hat{\rho}(\lambda)$ is spectral reflectance estimation of an object color.

$$
\begin{equation*}
D=\int \sum_{i=1}^{6}\{\hat{\rho}(\lambda)-\rho(\lambda)\}^{2} d \lambda \tag{5}
\end{equation*}
$$

The best estimation condition is given by Eq. 6.

$$
\begin{equation*}
\frac{\partial D}{\partial w}=0 \tag{6}
\end{equation*}
$$

Equation 7 is obtained by substituting Eq. 5 into Eq. 6 and implementing partial differentiation.

$$
\begin{equation*}
\int e_{i}(\lambda)\left\{\sum_{i=1}^{6} w_{i} \cdot e_{i}(\lambda)-\rho(\lambda)\right\} d \lambda=0 \tag{7}
\end{equation*}
$$

Eq. 7 can be written as Eq. 8.

$$
\begin{equation*}
\int e_{i}(\lambda)\left\{\sum_{i=1}^{6} w_{i} \cdot e_{i}(\lambda)\right\} d \lambda=\int e_{i}(\lambda) \rho(\lambda) d \lambda=b_{i} \tag{8}
\end{equation*}
$$

Matrix expression of Eq. 8 is obtained by Eq. 9.
$\left(\begin{array}{llllll}E_{1} \cdot E_{1} & E_{1} \cdot E_{2} & E_{1} \cdot E_{3} & E_{1} \cdot E_{4} & E_{1} \cdot E_{5} & E_{1} \cdot E_{6} \\ E_{2} \cdot E_{1} & E_{2} \cdot E_{2} & E_{2} \cdot E_{3} & E_{2} \cdot E_{4} & E_{2} \cdot E_{5} & E_{2} \cdot E_{6} \\ E_{3} \cdot E_{1} & E_{3} \cdot E_{2} & E_{3} \cdot E_{3} & E_{3} \cdot E_{4} & E_{3} \cdot E_{5} & E_{3} \cdot E_{6} \\ E_{4} \cdot E_{1} & E_{4} \cdot E_{2} & E_{4} \cdot E_{3} & E_{4} \cdot E_{4} & E_{4} \cdot E_{5} & E_{4} \cdot E_{6} \\ E_{5} \cdot E_{1} & E_{5} \cdot E_{2} & E_{5} \cdot E_{3} & E_{5} \cdot E_{4} & E_{5} \cdot E_{5} & E_{5} \cdot E_{6} \\ E_{6} \cdot E_{1} & E_{6} \cdot E_{2} & E_{6} \cdot E_{3} & E_{6} \cdot E_{4} & E_{6} \cdot E_{5} & E_{6} \cdot E_{6}\end{array}\right)\left(\begin{array}{l}w_{1} \\ w_{2} \\ w_{3} \\ w_{4} \\ w_{5} \\ w_{6}\end{array}\right)=\left(\begin{array}{l}b_{1} \\ b_{2} \\ b_{3} \\ b_{4} \\ b_{5} \\ b_{6}\end{array}\right)$
where, $b_{i}$ are constant when spectral reflectance of an object color $\rho(\lambda)$ was given.

For simplicity, modify Eq. 1 as Eq. 10.

$$
\left.\begin{array}{l}
E_{1}=\sin \left(\frac{1}{2} \pi x\right), E_{2}=\cos \left(\frac{1}{2} \pi x\right), E_{3}=\sin (\pi x)  \tag{10}\\
E_{4}=\cos \left(\frac{3}{2} \pi x\right), E_{5}=\sin (2 \pi x), E_{6}=\cos \left(\frac{5}{2} \pi x\right)
\end{array}\right\}
$$

where, $0 \leq x \leq 1$
The each element of left hand side of matrix in Eq. 10 can be calculated in the following manner.

$$
\begin{aligned}
& E_{1} \cdot E_{1}=\int_{0}^{\frac{\pi}{2}} \sin (x) \sin (x) d x, E_{1} \cdot E_{2}=\int_{0}^{\frac{\pi}{2}} \sin (x) \cos (x) d x \\
& E_{1} \cdot E_{3}=\int_{0}^{\frac{\pi}{2}} \sin (x) \sin (2 x) d x, E_{1} \cdot E_{4}=\int_{0}^{\frac{\pi}{2}} \sin (x) \cos (3 x) d x \\
& E_{1} \cdot E_{5}=\int_{0}^{\frac{\pi}{2}} \sin (x) \sin (4 x) d x, E_{1} \cdot E_{6}=\int_{0}^{\frac{\pi}{2}} \sin (x) \cos (5 x) d x \\
& E_{2} \cdot E_{1}=\int_{0}^{\frac{\pi}{2}} \cos (x) \sin (x) d x, E_{2} \cdot E_{2}=\int_{0}^{\frac{\pi}{2}} \cos (x) \cos (x) d x \\
& E_{2} \cdot E_{3}=\int_{0}^{\frac{\pi}{2}} \cos (x) \sin (2 x) d x, E_{2} \cdot E_{4}=\int_{0}^{\frac{\pi}{2}} \cos (x) \cos (3 x) d x \\
& E_{2} \cdot E_{5}=\int_{0}^{\frac{\pi}{2}} \cos (x) \sin (4 x) d x, E_{2} \cdot E_{6}=\int_{0}^{\frac{\pi}{2}} \cos (x) \cos (5 x) d x \\
& E_{3} \cdot E_{1}=\int_{0}^{\frac{\pi}{2}} \sin (2 x) \sin (x) d x, E_{3} \cdot E_{2}=\int_{0}^{\frac{\pi}{2}} \sin (2 x) \cos (x) d x \\
& E_{3} \cdot E_{3}=\int_{0}^{\frac{\pi}{2}} \sin (2 x) \sin (2 x) d x, E_{3} \cdot E_{4}=\int_{0}^{\frac{\pi}{2}} \sin (2 x) \cos (3 x) d x \\
& E_{3} \cdot E_{5}=\int_{0}^{\frac{\pi}{2}} \sin (2 x) \sin (4 x) d x, E_{3} \cdot E_{6}=\int_{0}^{\frac{\pi}{2}} \sin (2 x) \cos (5 x) d x \\
& E_{4} \cdot E_{1}=\int_{0}^{\frac{\pi}{2}} \cos (3 x) \sin (x) d x, E_{4} \cdot E_{2}=\int_{0}^{\frac{\pi}{2}} \cos (3 x) \cos (x) d x
\end{aligned}
$$

$$
\begin{aligned}
& E_{4} \cdot E_{3}=\int_{0}^{\frac{\pi}{2}} \cos (3 x) \sin (2 x) d x, E_{4} \cdot E_{4}=\int_{0}^{\frac{\pi}{2}} \cos (3 x) \cos (3 x) d x \\
& E_{4} \cdot E_{5}=\int_{0}^{\frac{\pi}{2}} \cos (3 x) \sin (4 x) d x, E_{4} \cdot E_{6}=\int_{0}^{\frac{\pi}{2}} \cos (3 x) \cos (5 x) d x \\
& E_{5} \cdot E_{1}=\int_{0}^{\frac{\pi}{2}} \sin (4 x) \sin (x) d x, E_{5} \cdot E_{2}=\int_{0}^{\frac{\pi}{2}} \sin (4 x) \cos (x) d x \\
& E_{5} \cdot E_{3}=\int_{0}^{\frac{\pi}{2}} \sin (4 x) \sin (2 x) d x, E_{5} \cdot E_{4}=\int_{0}^{\frac{\pi}{2}} \sin (4 x) \cos (3 x) d x \\
& E_{5} \cdot E_{5}=\int_{0}^{\frac{\pi}{2}} \sin (4 x) \sin (4 x) d x, E_{5} \cdot E_{6}=\int_{0}^{\frac{\pi}{2}} \sin (4 x) \cos (5 x) d x \\
& E_{6} \cdot E_{1}=\int_{0}^{\frac{\pi}{2}} \cos (5 x) \sin (x) d x, E_{6} \cdot E_{2}=\int_{0}^{\frac{\pi}{2}} \cos (5 x) \cos (x) d x \\
& E_{6} \cdot E_{3}=\int_{0}^{\frac{\pi}{2}} \cos (5 x) \sin (2 x) d x, E_{6} \cdot E_{4}=\int_{0}^{\frac{\pi}{2}} \cos (5 x) \cos (3 x) d x \\
& E_{6} \cdot E_{5}=\int_{0}^{\frac{\pi}{2}} \cos (5 x) \sin (4 x) d x, E_{6} \cdot E_{6}=\int_{0}^{\frac{\pi}{2}} \cos (5 x) \cos (5 x) d x
\end{aligned}
$$

Therefore Eq. 9 can be re-written as Eq. 11.
$\left(\begin{array}{cccccc}\pi / 4 & 1 / 2 & 2 / 3 & -1 / 2 & -4 / 15 & 1 / 6 \\ 1 / 2 & \pi / 4 & 2 / 3 & 0 & -4 / 15 & 0 \\ 2 / 3 & 2 / 3 & \pi / 4 & -2 / 5 & 0 & 2 / 21 \\ -1 / 2 & 0 & -2 / 5 & \pi / 4 & 4 / 7 & 0 \\ -4 / 15 & -4 / 15 & 0 & 4 / 7 & \pi / 4 & 4 / 9 \\ 1 / 6 & 0 & 2 / 21 & 0 & 4 / 9 & \pi / 4\end{array}\right) \cdot\left(\begin{array}{l}w_{1} \\ w_{2} \\ w_{3} \\ w_{4} \\ w_{5} \\ w_{6}\end{array}\right)=\left(\begin{array}{l}b_{1} \\ b_{2} \\ b_{3} \\ b_{4} \\ b_{5} \\ b_{6}\end{array}\right)$

Then, 6 weighting factors $w_{1} \sim w_{6}$ are obtained by Eq. 12.

$$
\left(\begin{array}{l}
w_{1}  \tag{12}\\
w_{2} \\
w_{3} \\
w_{4} \\
w_{5} \\
w_{6}
\end{array}\right)=\left(\begin{array}{cccccc}
\pi / 4 & 1 / 2 & 2 / 3 & -1 / 2 & -4 / 15 & 1 / 6 \\
1 / 2 & \pi / 4 & 2 / 3 & 0 & -4 / 15 & 0 \\
2 / 3 & 2 / 3 & \pi / 4 & -2 / 5 & 0 & 2 / 21 \\
-1 / 2 & 0 & -2 / 5 & \pi / 4 & 4 / 7 & 0 \\
-4 / 15 & -4 / 15 & 0 & 4 / 7 & \pi / 4 & 4 / 9 \\
1 / 6 & 0 & 2 / 21 & 0 & 4 / 9 & \pi / 4
\end{array}\right)^{-1} \cdot\left(\begin{array}{l}
b_{1} \\
b_{2} \\
b_{3} \\
b_{4} \\
b_{5} \\
b_{6}
\end{array}\right)
$$

The Straightforward calculation of multiple regression coefficients is given by Eq. 13.

$$
\left(\begin{array}{l}
w_{1}  \tag{13}\\
w_{2} \\
w_{3} \\
w_{4} \\
w_{5} \\
w_{6}
\end{array}\right)=\left(\begin{array}{cccccc}
0.45727 & -4.9302 & 6.6638 & 5.8619 & -2.9898 & -0.97809 \\
-4.9302 & 580.18 & -811.41 & -655.31 & 327.70 & 87.578 \\
6.6638 & -811.41 & 1135.4 & 917.16 & -459.05 & -122.80 \\
5.8619 & -655.31 & 917.16 & 742.27 & -372.25 & -100.13 \\
-2.9898 & 327.70 & 459.05 & -372.25 & 187.33 & 50.719 \\
-0.97809 & 87.578 & -122.80 & -100.13 & 50.719 & 14.021
\end{array}\right)^{-1}\left(\begin{array}{l}
b_{1} \\
b_{2} \\
b_{3} \\
b_{4} \\
b_{5} \\
b_{6}
\end{array}\right)
$$

## Straightforward decoding by three dimensional $1^{\text {st }}$ order equations

The matrix expression of Eq. 4 is given by Eq. 14,
$\left(\begin{array}{c}X-\hat{X} \\ Y-\hat{Y} \\ Z-\hat{Z}\end{array}\right)=\left(\begin{array}{lll}S x_{4} & S x_{5} & S x_{6} \\ S y_{4} & S y_{5} & S y_{6} \\ S z_{4} & S z_{5} & S z_{6}\end{array}\right) \bullet\left(\begin{array}{c}w_{4} \\ w_{5} \\ w_{6}\end{array}\right)$
where,
$S x_{i}=\int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{x}(\lambda) d \lambda$
$S y_{i}=\int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{y}(\lambda) d \lambda$
$S z_{i}=\int e_{i}(\lambda) \cdot E(\lambda) \cdot \bar{z}(\lambda) d \lambda$
( $i=4,5,6$ )
3 weighting factors $w_{4} w_{5} w_{6}$ are obtained by Eq. 15.

$$
\left(\begin{array}{l}
w_{4}  \tag{15}\\
w_{5} \\
w_{6}
\end{array}\right)=\left(\begin{array}{lll}
S_{4} & S_{5} & S_{6} \\
S_{4} & S_{5} & S_{6} \\
S_{4} & S_{5} & S_{6}
\end{array}\right)^{-1} \cdot\left(\begin{array}{c}
X-\hat{X} \\
Y-\hat{Y} \\
Z-\hat{Z}
\end{array}\right)
$$

Above $S_{i}$ is a function of illuminant. So, the matrices for D65, D50 and BBR4K (black body radiation of 4000 K ) are shown bellow.

$$
\begin{aligned}
& \left(\begin{array}{lll}
S_{4} & S_{5} & S_{6} \\
S_{4} & S_{5} & S_{6} \\
S_{4} & S_{5} & S_{6}
\end{array}\right)^{-1}=\left(\begin{array}{ccc}
-0.01262 & -0.01118 & -0.00600 \\
0.00628 & 0.00990 & 0.01473 \\
0.02250 & -0.01186 & 0.00740
\end{array}\right) \\
& \left(\begin{array}{lll}
S_{4} & S_{5} & S_{6} \\
S_{4} & S_{5} & S_{6} \\
S_{4} & S_{5} & S_{6}
\end{array}\right)_{D 50}=\left(\begin{array}{ccc}
-0.01063 & -0.01158 & -0.00914 \\
0.00566 & 0.00921 & 0.00202 \\
0.02169 & -0.01298 & 0.01101
\end{array}\right) \\
& \left(\begin{array}{lll}
S_{4} & S_{5} & S_{6} \\
S_{4} & S_{5} & S_{6} \\
S_{4} & S_{5} & S_{6}
\end{array}\right)_{\text {BBR4K}}=\left(\begin{array}{ccc}
-0.00777 & -0.01330 & -0.01298 \\
0.00376 & 0.01049 & 0.02692 \\
0.02028 & -0.01317 & 0.01518
\end{array}\right)
\end{aligned}
$$

## Spectral encoding/decoding result of real multi-spectral images

The real multi-spectral images as shown in Fig. 2, copyrighted by Miyake Laboratory, Chiba University, were in 16bits 5 channels TIFF format.

The spectral reflectance curve of each pixel of an image $\rho_{c u}(\lambda)$ was calculated from 5 channels image data using its own multiple regression formula. Then the CIEXYZ and $s R G B$ values were calculated from the spectral reflectance curve $\rho_{c u}(\lambda)$. The objective of LabRGB, here, is to encoding/decoding the above spectral reflectance curve $\rho_{c u}(\lambda)$ to obtain $\hat{\rho}(\lambda)$ with a minimum spectral and colorimetric estimation error.

Encoding is done by substituting $\rho_{c u}(\lambda)$ into Eq. 8 to obtain $b_{1} \sim b_{6}$, substituting $b_{1} \sim b_{6}$ into Eq. 13 and calculate CIEXYZ and CIELAB values from $\rho_{c u}(\lambda)$ to obtain $L a b R G B$ encoding values $L^{*} a^{*} b^{*} w_{1} w_{2} w_{3}$.

Decoding is done by Eq. 3 and Eq. 14 to obtain $w_{4} w_{5} w_{6}$, then substituting $w_{4} w_{5} w_{6}$ and the last three LabRGB encoding values $w_{1} w_{2} w_{3}$ into Eq. 2 to obtain $\hat{\rho}(\lambda)$.

The LabRGB encoding/decoding images were shown in Fig. 3. D50 was used in encoding/decoding. The original images in Fig. 2 and the LabRGB encoding/decoding images in Fig. 3 look almost the same and cannot tell the
difference by human eye. The colorimetric and spectral accuracies were shown in Table 1 and Table 2 respectively. Surprisingly colorimetric and spectral estimation errors were small for all images.

Figure. 4 shows the local worst area in P1 image in terms of $\Delta E a b$. More than $2 \Delta E_{a b}$ are indicated as white color. For example, the saturated colors in right top of P1 image were more than $2 \Delta E_{a b}$. Therefore the encoding/decoding using illuminant D65 observed by illuminant A is the worst case in P1 image. The colorimetric accuracies show the same trend in Table 1.

Figures.5, 6 and 7 show all P1~P8 images observed by illuminant A with the encoding/decoding using illuminant D65, D50 and BBR4K respectively. Illuminant D50 and BBR4K are much better than illuminant D65, but the difference between illuminant D50 and BBR4K are not so significant. Thus, for encoding/decoding, illuminant D50 is an appropriate choice to minimize $\Delta E_{a b}$ for the local worst area.

It should be noted that we can eliminate color estimation error completely ( $\Delta E_{a b}=0$ ), if we choose the encoding/decoding illuminant same as the observation illuminant. It can be said by $L a b R G B$ definition.

Table 1. Comparison of the colorimetric estimation errors (standard deviation of the $764 \times 508$ pixel $\Delta E a b$ in each image)

| Image name | LabRGB encoding/decoding illuminants | Observation illuminants |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  | D65 | D50 | A |
| P1 | BBR4K | 0.2932 | 0.2187 | 0.2748 |
|  | D50 | 0.1647 | 0 | 0.4028 |
|  | D65 | 0 | 0.1620 | 0.5296 |
| P2 | BBR4K | 0.1281 | 0.0980 | 0.1228 |
|  | D50 | 0.0742 | 0 | 0.1722 |
|  | D65 | 0 | 0.0748 | 0.2289 |
| P3 | BBR4K | 0.2790 | 0.2277 | 0.2877 |
|  | D50 | 0.1452 | 0 | 0.4156 |
|  | D65 | 0 | 0.1529 | 0.5343 |
| P4 | BBR4K | 0.4138 | 0.3572 | 0.4667 |
|  | D50 | 0.2899 | 0 | 0.5702 |
|  | D65 | 0 | 0.2936 | 0.7871 |
| P5 | BBR4K | 0.3917 | 0.3252 | 0.4186 |
|  | D50 | 0.2473 | 0 | 0.5490 |
|  | D65 | 0 | 0.2521 | 0.7401 |
| P6 | BBR4K | 0.4737 | 0.3613 | 0.7141 |
|  | D50 | 0.4423 | 0 | 0.7505 |
|  | D65 | 0 | 0.4419 | 1.1115 |
| P7 | BBR4K | 0.2335 | 0.1816 | 0.1771 |
|  | D50 | 0.1071 | 0 | 0.2705 |
|  | D65 | 0 | 0.1102 | 0.3476 |
| P8 | BBR4K | 0.3749 | 0.3408 | 0.5382 |
|  | D50 | 0.2911 | 0 | 0.6627 |
|  | D65 | 0 | 0.2999 | 0.8872 |



Figure 2. The original images, copyrighted by Miyake Laboratory, Chiba University, were used in the encoding/decoding test. The original images were name as P1 ~ P8 and are located from top to bottom respectively. The pictures observed by D65, D50 and A located from left to right.


Figure 3. The images after LabRGB encoding/decoding took place. In this case, illuminant D50 was used in LabRGB encoding/decoding. The images were P1 ~ P8 located from top to bottom respectively. The pictures observed by D65, D50 and A located from left to right.


Figure 4. LabRGB encoding/decoding images P1. The pictures are for encoding/decoding using illuminants D65, D50 and A located from left to right. The pictures observed by D65, D50 and A located from top to bottom.

More than $2 \Delta E_{a b}$ are indicated as white color.


Figure 5. LabRGB encoding/decoding images P1 through P8. The pictures are for encoding/decoding using illuminant D65 observed by illuminant A. More than $2 \Delta E_{a b}$ are indicated as white color.

Table2. Comparison of the spectral estimation errors (standard deviation of the 31 wavelength sampling points $\times 764 \times 508$ pixel $\Delta \rho$ in each image)

| Image name | LabRGB encoding/decoding <br> illuminants |  |  |
| :---: | :---: | :---: | :---: |
|  | BBR4K | D50 | D65 |
|  | 0.005975 | 0.006022 |  |
| P2 | 0.005935 | 0.005972 | 0.005981 |
| P3 | 0.013721 | 0.013969 | 0.014044 |
| P4 | 0.008509 | 0.008642 | 0.008678 |
| P5 | 0.007921 | 0.008072 | 0.008114 |
| P6 | 0.009485 | 0.009574 | 0.009583 |
| P7 | 0.004600 | 0.004666 | 0.004686 |
| P8 | 0.021393 | 0.021711 | 0.021791 |



Figure 6. LabRGB encoding/decoding images P1 through P8. The pictures are for encoding/decoding using illuminant D50 observed by illuminant A. More
than $2 \Delta E_{a b}$ are indicated as white color.


Figure 7. LabRGB encoding/decoding images P1 through P8. The pictures are for encoding/decoding using illuminant BBR4K observed by illuminant A. More than $2 \Delta E_{a b}$ are indicated as white color.

## Conclusion

The spectral encoding/decoding method, called $L a b R G B$ derivation was described in detail including fast straightforward calculation of encoding/decoding calculation. LabRGB was applied to 8 different types of multi-spectral images to evaluate the encoding/decoding characteristics. LabRGB performed well in encoding/decoding various types of multi-spectral images and confirmed to be suitable for practical use.

Future plan is to accumulate $\operatorname{LabRGB}$ test data to multi-spectral imaging systems and continue a performance evaluation.

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## Biography

Fumio Nakaya received his B.S degree in Mechanical engineering from Keio University in Japan in 1976. Since 1976 he has worked in research and development divisions at Fuji Xerox Co., Ltd in Kanagawa, Japan. His work has primarily focused on image quality and image quality design, including microscopic image structure for high quality color image using dry toner, color management in multimedia equipment and systems. He is a member of the IS\&T and the Institute of Image Information and Television Engineers.

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