

An Effect of Color Channel Replacement

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Abstract

This paper investigates the effect of color channel replacement with histogram equalization. The histogram equalization is broadly employed for contrast intensification in a diverse of consumer electronics applications for its straightforward function. The proposed work is consists of seven steps, original given image, channel separation: color image to three gray scale image, color channel replacement, HEM on individually separated image, histogram modification in each color channel, reconstruct color images, and image display. Simulation results with test images are provided to compare its PSNR and MSE performance.

Keywords: *Image, histogram equalization, color image, color replacement*

1. Introduction

Histogram equalization method (HEM) is an important field in image processing for both human and computer vision. HEM changes original images in different ways, and this application includes photographs, photochemical photographs, and illustrations. The higher quality of images enables operators to manage state-of-art approaches for image enhancement [1-3]. HEM adjusts pixel intensities and the histogram of the resulting image becomes uniform.

HEM is broadly adopted in many applications including speech recognition, texture synthesis, medical image processing, and many other video processing applications [4-8]. There have been many applications to develop HEM [9]. Most of them employed straightforward linear or nonlinear gray level transformation functions. Some others have been using more complex functions or features such as edge or colors [10-12]. By all means, HEM is the most usually adopted approach due to its plainness and comparatively better performance on almost all kinds of images.

Generally speaking, HEM levels the density distribution of the derived images and increases the contrast of the image as a consequence. It is because HEM has an influence of dynamic range extending. However, despite of its better performance in intensifying contrasts of a provided image, it is hardly employed in consumer electronics applications such as television due to its straightforward usage of HEM, and its effect may affect the original brightness of an input image and eventually spoil the subjective quality.

In this paper, we used global HEM. The global HEM uses the histogram information of the whole input images for its transformation function. We assume gray scale image and color image are 1-D and 3-D approach as color image has three color channels [13]. The rest of this paper is composed as follows. In Section 2, we explain the proposed method. Section 3 shows simulation results and conclusion remarks are shown in Section 4.

2. Proposed method

Histogram equation method (HEM) intensifies the images contrast by changing the intensity values of given images. Sometimes it can be used for indexed colormap image values. The HEM normally enhances the universal contrast of images. It is useful when the adopted image data is outlined with close contrast intensities. In this case, the strengths can be adjusted by changing the intensities of the histogram. This is the main idea that one may enhance low contrast image to high contrast image. One of the main benefits of HEM is that HEM is a simple approach. Therefore, if one knows the histogram equalization function, then the original low contrast image can be regained perfectly. Let us assume a discontinuous grayscale image x , and suppose k_i be the number of events of gray level i . The probability of an event of a pixel of level i in the original image is

$$\begin{aligned} pr_g(i) &= pr(g=i) \\ &= \frac{k_i}{k}, \end{aligned} \quad (1)$$

where $0 < i < L$. We assume L is the total number of gray levels in the image. Since we assumed g as grey image, L is 256. Parameter k is the number of pixels in an image. Then, the cumulative distribution function is,

$$CDF_g(i) = \sum_{j=0}^i pr_g(j). \quad (2)$$

Here, CDF is cumulative distribution function which depicts the probability that a real-valued random variable M with a granted probability distribution is discovered at a value equal to or less than m . Now, to map the values back into their original range, the following simple transformation needs to be applied on the result:

$$h' = h \cdot (g_{\max} - g_{\min}) + g_{\min}. \quad (3)$$

This is the global histogram equalization and we use this approach in the proposed method.

Although above HEM is for the grayscale image, it can also be employed on color images by applying the HEM individually to red, green and blue channels of color image. The study compares the effect of HEM on color replaced images. Figure 1 shows the block diagram of the proposed method. The proposed method has seven steps.

- Step 1: Original given image
- Step 2: Channel separation: Color image to three gray scale image
- Step 3: Color channel replacement
- Step 4: HEM on individually separated image
- Step 5: Histogram modification in each color channel
- Step 6: Reconstruct color images
- Step 7: Image display

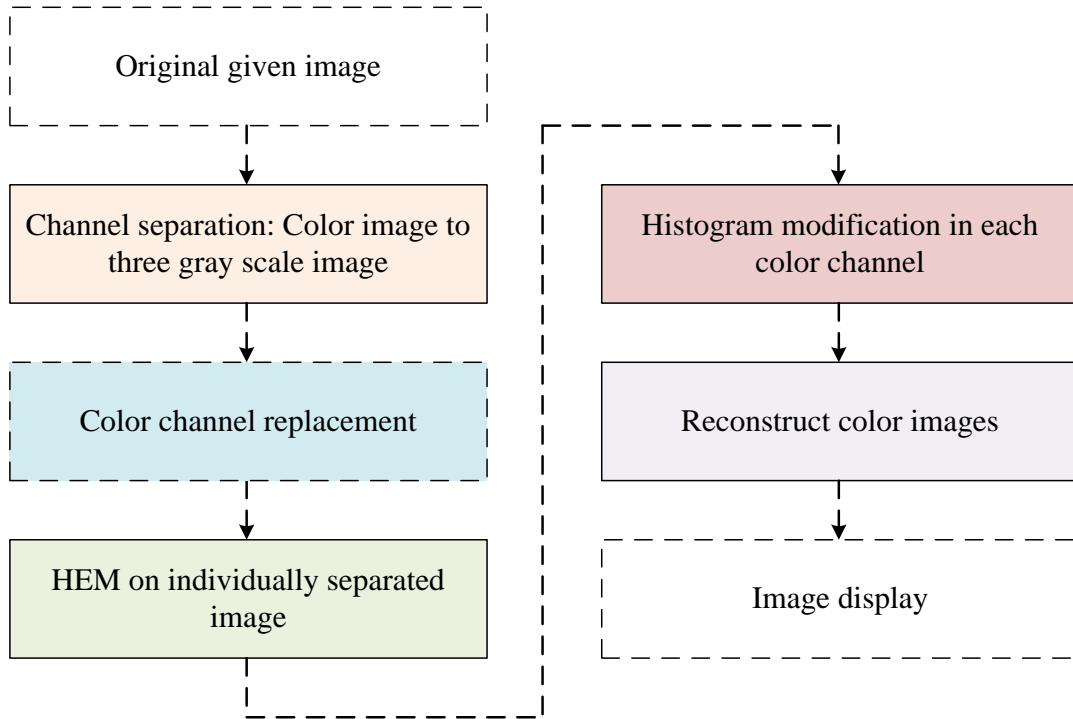
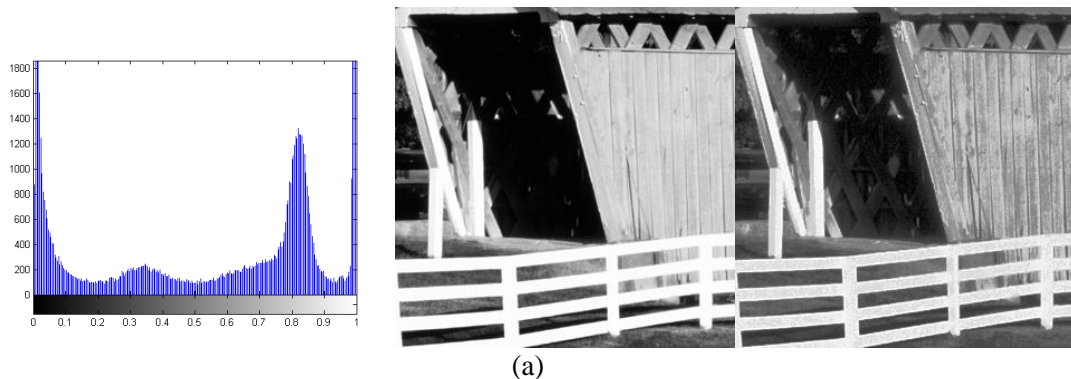


Figure 1. Flowchart of the entropy method

Figure 2 shows the histogram of #18 McM image (*left*), original greyscale image of each color channel, and its histogram equalized image. The rightest images of Figure 2 show that the employed HEM gives a meaningful enhancement in image contrast. Figures 2(a), 2(b), and 2(c) show the red, green, and blue channels, respectively. As we can see HEM method applied image exceeds the other original image by intensifying the contrast very well. It is noted that the serious side effects such as checkerboard effects or washed out appearance were not generated during this process.



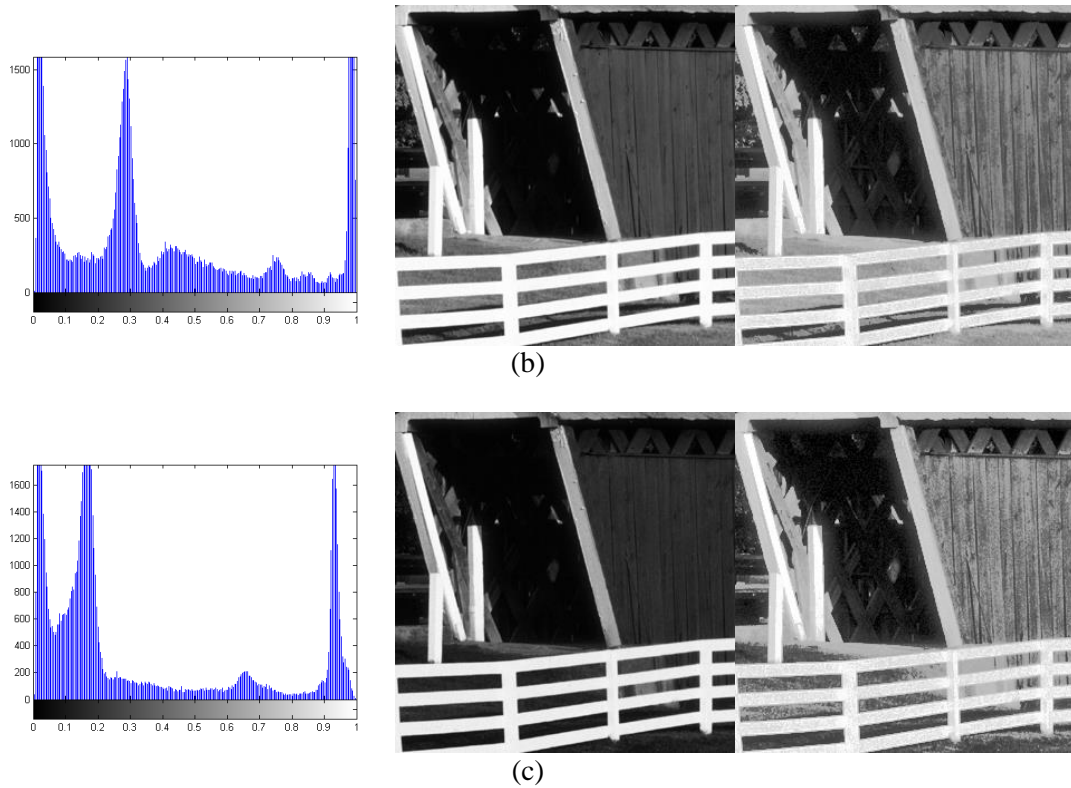


Figure 2. Histogram of #18 McM image and its original color channel and its histogram equalized image: (a) red, (b) green, and (c) blue. Note that the leftist image shows the original histogram and the rightest image is with specified histogram

3. Simulation results

The simulation results from introduced method were conducted on several images and assessed the performance. The simulation was conducted on 18 McM test images, which are shown in Figure 3.



Figure 3. Eighteen test image set: McM images #1 to #18

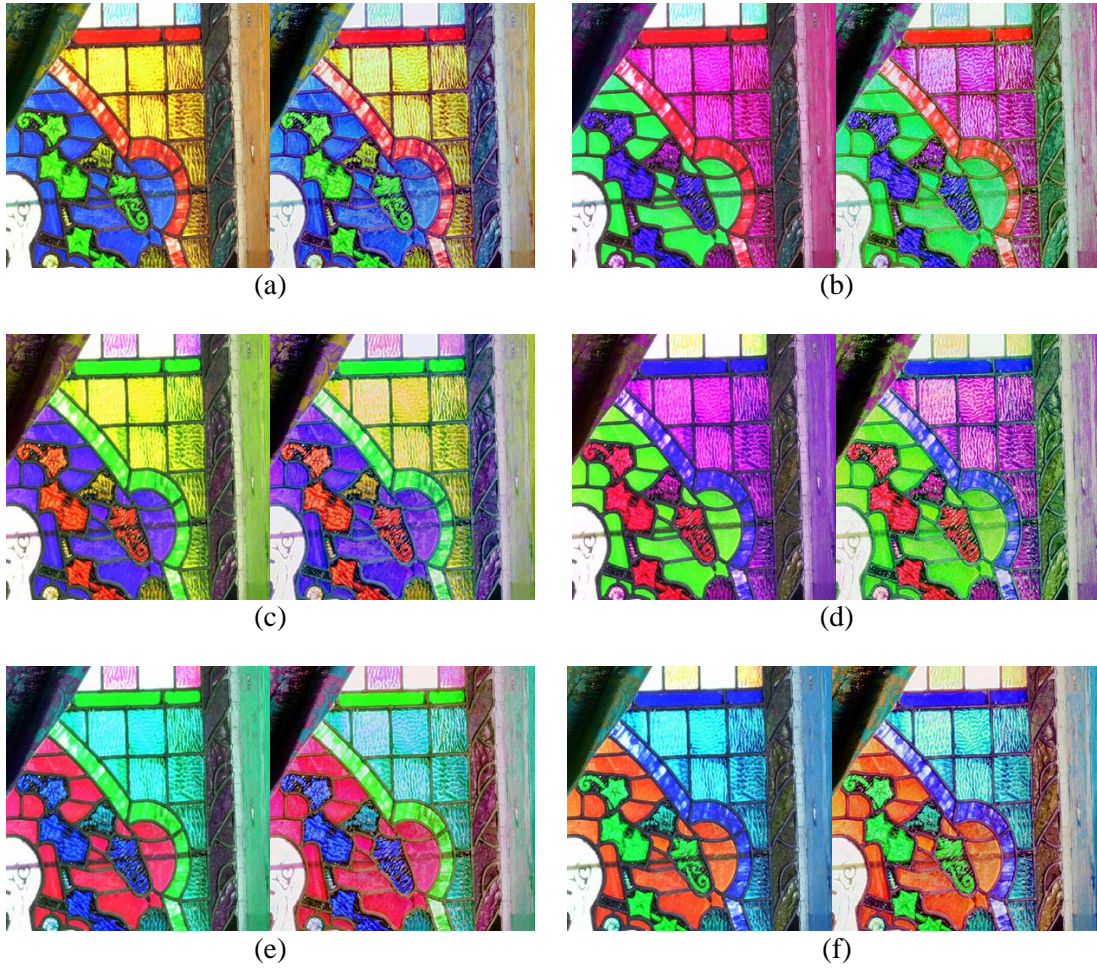


Figure 4. Result images of McM #1 image with different combination of R, G, B channels and its histogram equalized results: (a) RGB, (b) RBG, (c) GRB, (d) GBR, (e) BRG, and (f) BGR

Figure 4(a) shows the RGB original image along with simulation results from HEM on McM image #1. In the same manner, Figures 4(b-f) shows different combination of red, green and blue color channels along with simulation results from HEM. They are RBG, GRB, GBR, BRG, and BGR, respectively.

In the same manner, we obtained two other examples with McM images #5 and #14, which are shown in Figures 5-6.



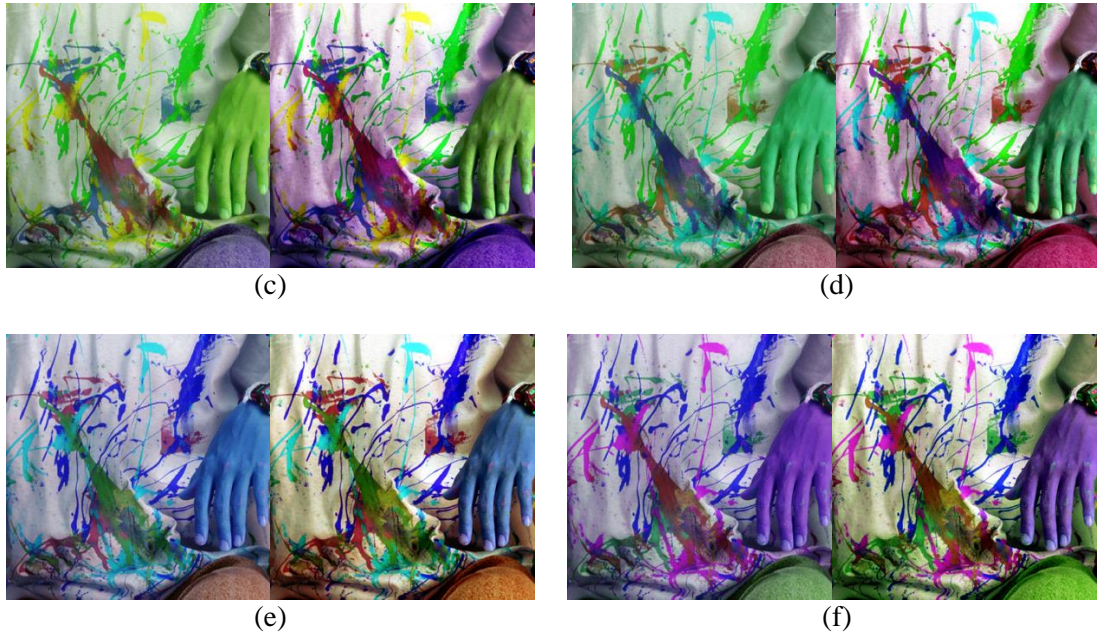
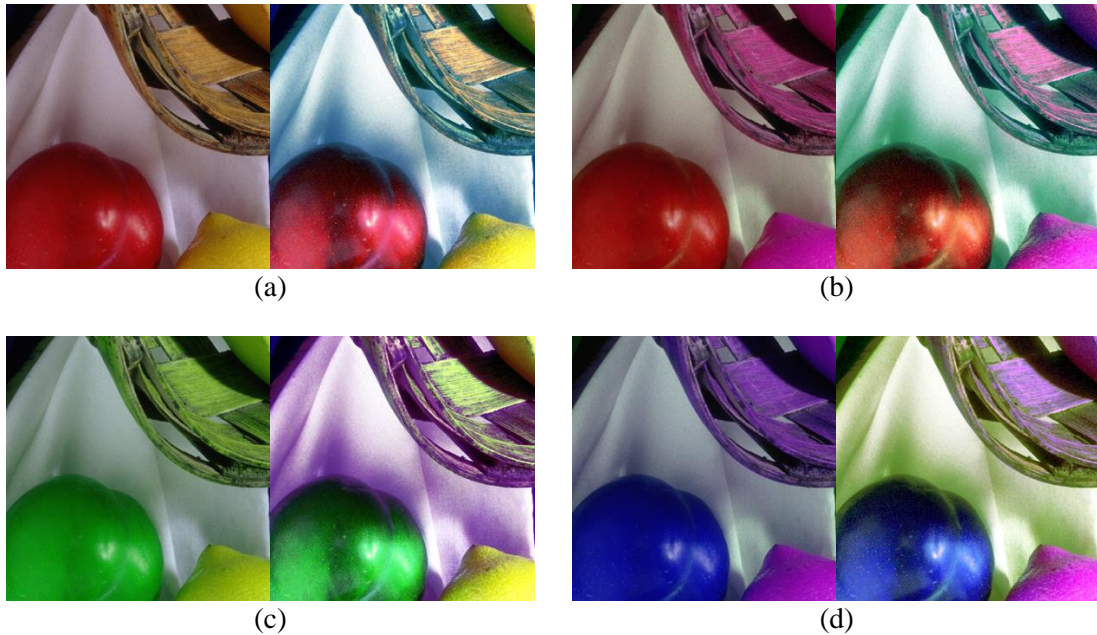


Figure 5. Result images of McM #5 image with different combination of R, G, B channels and its histogram equalized results: (a) RGB, (b) RBG, (c) GRB, (d) GBR, (e) BRG, and (f) BGR

Table 1 shows the PSNR comparison for 18 McM images (#1 to #18 images) with replaced red, green, and blue combinations. Figures 7 and 8 show the performance comparisons in CPSNR (dB) and CMSE for replaced red, green, and blue combinations.



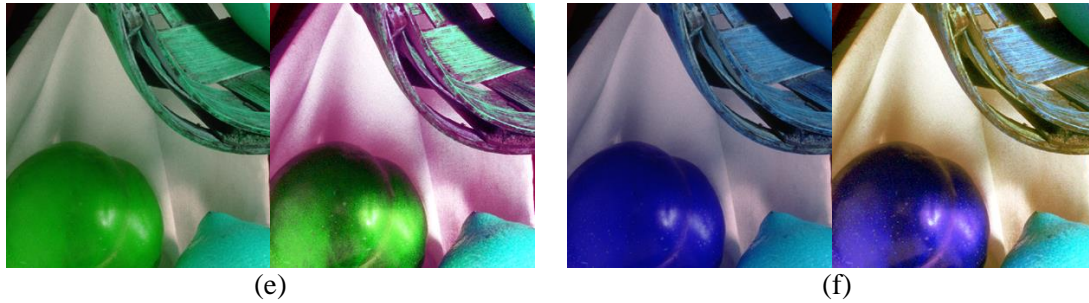


Figure 6. Result images of McM #14 image with different combination of R, G, B channels and its histogram equalized results: (a) RGB, (b) RBG, (c) GRB, (d) GBR, (e) BRG, and (f) BGR

Table 1. PSNR comparison for 18 McM images (1st to 18^h images) with varying red, green, and blue combinations

	R channel			G channel			B channel		
	RGB	GBR	BRG	RGB	GBR	BRG	RGB	GBR	BRG
1	23.604	11.053	7.449	23.924	8.944	11.425	14.846	6.793	8.150
2	17.372	14.247	11.243	14.549	12.273	13.616	16.409	10.975	12.995
3	17.515	14.827	15.511	16.898	15.519	14.545	11.750	11.145	11.425
4	13.125	13.029	13.082	11.171	11.078	11.018	13.401	13.213	13.222
5	16.410	12.583	11.286	21.431	15.207	13.914	20.710	11.898	14.819
6	16.973	14.690	10.885	12.513	10.564	11.506	10.312	8.783	9.575
7	20.105	12.217	12.088	15.282	15.204	11.380	14.434	11.050	14.427
8	8.806	8.172	8.155	8.329	8.278	8.225	7.818	7.739	7.810
9	20.526	11.975	12.254	16.011	14.729	11.754	11.062	9.405	10.598
10	21.850	10.374	9.446	8.837	7.523	7.208	7.827	6.654	7.221
11	18.814	8.662	8.387	8.528	6.914	6.378	6.203	5.272	6.059
12	13.992	12.442	12.081	13.142	12.792	11.496	9.224	8.552	9.121
13	12.446	11.906	11.557	13.698	13.392	12.577	11.385	10.471	11.165
14	16.078	13.023	12.407	13.216	11.863	10.598	12.085	9.623	11.474
15	21.135	13.353	12.807	6.522	6.340	5.670	6.265	5.639	6.382
16	14.379	14.090	7.076	20.607	8.535	19.404	6.003	5.651	5.681
17	18.022	7.444	7.278	8.053	6.594	5.963	8.412	6.233	6.091
18	17.844	13.162	14.692	17.687	16.427	13.425	12.281	11.160	12.314
Avg.	17.166	12.069	10.982	13.911	11.232	11.117	11.135	8.903	9.918

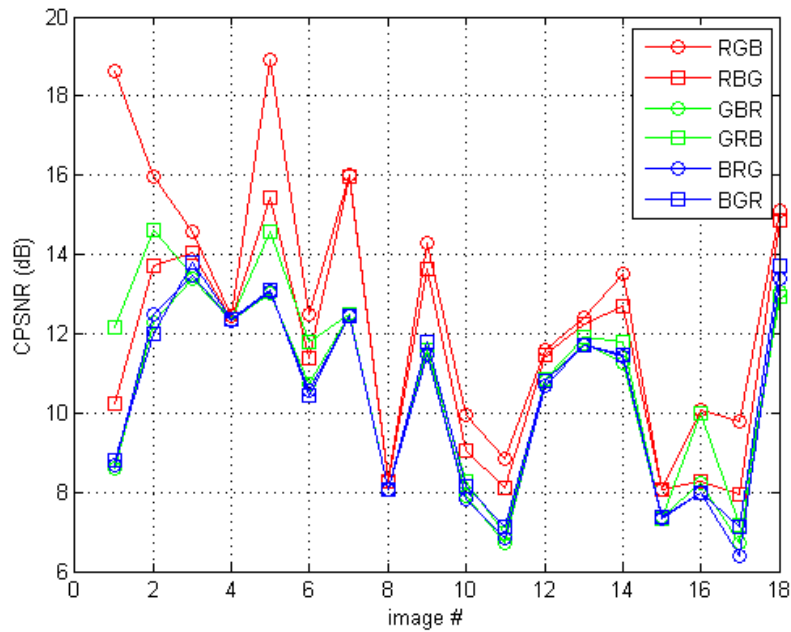


Figure 7. Performance comparison in CPSNR (dB): different combination of red, green, and blue combinations

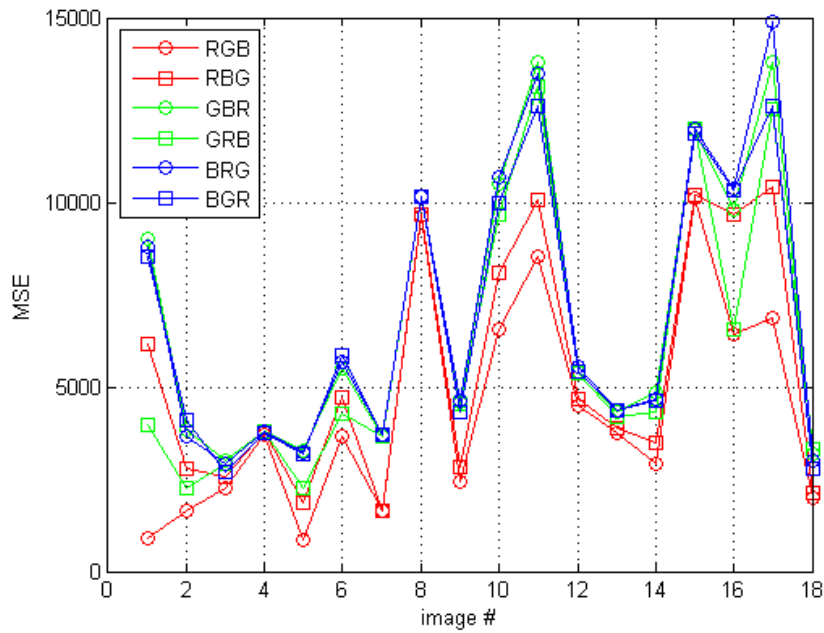


Figure 8. Performance comparison in CMSE: different combination of red, green, and blue combinations

4. Conclusions

This paper presented the effect of color channel replacement with histogram equalization. The proposed system is composed of seven stages, original given image, channel separation: color image to three gray scale image, color channel replacement, HEM on individually separated image, histogram modification in each color channel, reconstruct color images, and image display. The experimental results display the resultant images, and provide PSNR and MSE performance on 18 McM imageset.

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