

# Image Enhancement Using Weighted Bi-Histogram Equalization

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**Abstract**—Image enhancement is one of using in various digital signal processing areas. Advances in microcontrollers, microcomputers and computers have developed traditional algorithms in order to improve the quality of the resulting image and have implied many avenues to the design of new innovations using various techniques. This paper proposes contrast enhancement using weighted bi-histogram equalization based on distributed area ratio. Moreover, this technique must use a weighted factor which is calculated by the ratio of the histogram distribution. Likewise, an original image will be equalized by the modification of the probability density function of the gray levels. As result of the experiment, the contrast of resulting image is improved for implement and human perception – as well as also reduces the absolute mean brightness error (AMBE) than the traditional technique of image enhancement.

**Keywords**— Histogram Equalization, Weighted-Histogram Equalization, Area Ratio, Brightness Preservation, Tenengrad, Absolute Mean Brightness Error

## I. INTRODUCTION

Histogram Equalization (HE) is one of traditional techniques to increase the contrast of the image in digital image processing area. This technique is widely used to improve an image because it is easy to process and is simply. On the other hand, HE technique will produce the resulting image that is over-enhancement (also known as the quantum jump [1]). For instance, if an original image consists of most dark pixels, the resulting image will be extremely washed-out as shown in Fig.1. If an original one consists of most light pixels, the unwanted information will be cut-in as shown in Fig.2. In two cases, the resulting image will absolutely lost important information during histogram distribution.

Accordingly, there are many research [2], [3], [4] that present to preserve the brightness error. Thus, this proposed method aims to reduce its error using the weighted bi-histogram equalization which deals with problems.

## II. HISTOGRAM EQUALIZATION

Define  $\mathbf{X} = \{X(i, j)\}$  is an original image of discrete gray levels as  $\{X_0, X_1, \dots, X_{L-1}\}$ , where  $X(i, j)$  is an original image intensity of the pixels  $(i, j)$  and  $X(i, j) \in \{X_0, X_1, \dots, X_{L-1}\}$ .

$L$

Step 1: calculate the probability density function (PDF) is determined

$$p(l) = \frac{n_l}{N} ; l=0,1,\dots,L-1 \quad (1)$$

where  $n_l$  is the  $l$ -th of the gray level in the original image  $\mathbf{X}$  and  $N$  is the total number of pixel in the original image.

Step 2: calculate the cumulative density function (CDF) is determined as

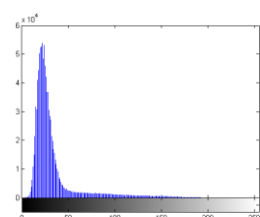
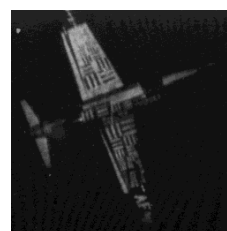
$$c(l) = \sum_{i=1}^{L-1} p(l) \quad (2)$$

Note  $0 \leq c(l) \leq 1$

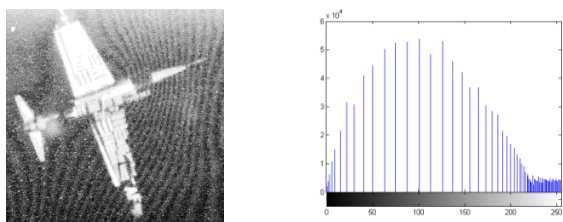
Step 3: calculate the new gray level of the resulting image using the mapping function is determined as

$$f(l) = X_0 + (X_{L-1} - X_0) \cdot c(l) \quad (3)$$

where  $X_0$  is the starting gray level of the resulting image.  $X_{L-1}$  is the ending gray level of the resulting image. Typically,  $X_0$  and  $X_{L-1}$  are defined as 0 and 255, respectively.

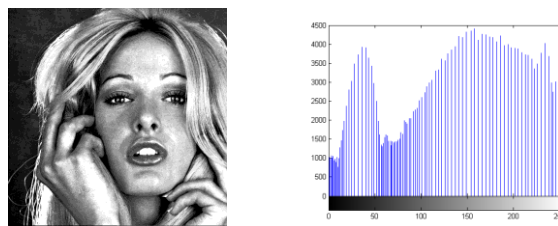


(a)



(b)

Fig. 1 image that consists of most dark pixels  
 (a) original image and its corresponding histogram  
 (b) resulting image using HE and its corresponding histogram



(b)

Fig.2 image that consists of most light pixels  
 (a) original image and its corresponding histogram  
 (b) resulting image using HE and its corresponding histogram

Step 4: represent the new gray level from step 3 into the original image. Then, the resulting image can be defined as

$$Y = \{f(X(i, j)) | \forall x(i, j) \in X\} \quad (4)$$

Typically, the histogram Equalization starting from  $X_0 = 0$  to  $X_{L-1} = 255$  is widely known as “Global Histogram Equalization (GHE)”.

HE algorithm always makes the high contrast of a resulting image to emphasise some information in an image however it also affects brightness errors in resulting one. [5]

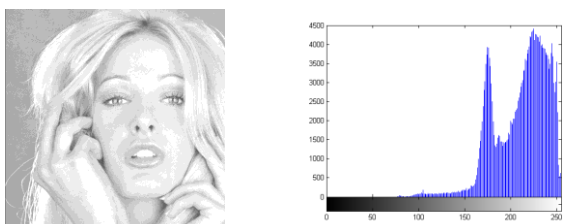
### III. BRIGHTNESS PRESERVATION USING WEIGHTED BI-HISTOGRAM EQUALIZATION

Define  $X_{ori} = \{X_{ori}(i, j)\}$  is an original image of discrete gray level as  $\{X_0, X_1, \dots, X_{L-1}\}$ , where  $X_{ori}(i, j)$  is an intensity of the pixel  $(i, j)$  and  $X_{ori}(i, j)$  is the number of  $\{X_0, X_1, \dots, X_{L-1}\}$

Step 1: Calculate the PDF is determined as

$$p(a) = \frac{n_a}{N}; a = 0, 1, \dots, L-1 \quad (5)$$

where  $a$  is the  $a$ -th of the gray level in the image  $X$  and  $n$  is the total number in each bin of the gray level.  $N$  is the total number of the whole image.



(a)

Step 2: Calculate the CDF is determined as

$$c(a) = \sum_{a=1}^{L-1} p(a) \quad (6)$$

Step 3: Determine the threshold level for dividing the whole histogram of the original image into two sections is determined as

$$\tau = \lim_{c \rightarrow 0.5} f(X) \quad (7)$$

the  $\tau$ -th gray level is used for dividing the whole histogram of the original image into two sections whereas about 50-percent of pixels are originated from the first region – as well as remaining pixels are from the second region.

Step 4: Calculate weighting factors are determined as

$$w_1 = \frac{s_0}{2\tau} \quad (8)$$

$$w_2 = \frac{s_1 - \tau}{256} + \frac{1}{2} \quad (9)$$

where  $s_0$  is the starting gray level of the original histogram.  $s_1$  is the ending gray level of the original histogram as shown in Fig.3

Define  $X_1 = \{x_1(i, j)\}$  is the first region of the original histogram started from  $L$  discrete gray level as  $\{s_0, s_{0+1}, \dots, \tau\}$ .  $X_2 = \{x_2(i, j)\}$  is the second region of the original histogram started from  $L$  discrete gray level as  $\{\tau + 1, \tau + 2, \dots, L-1\}$ .

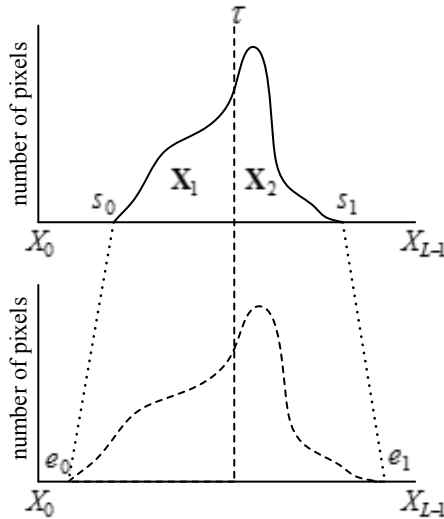


Fig.3 The concept of our proposed method

Step 5: Calculate PDF in each region is determined as

$$p_1(y) = \frac{n_y}{N_1}; y = 0, 1, \dots, \tau \quad (10)$$

$$p_2(z) = \frac{n_z}{N_2}; z = \tau + 1, \tau + 2, \dots, L - 1 \quad (11)$$

where  $n_y$  is the  $y$ -th of the gray level in the image  $X_1$  started from  $X_0$ -th gray level to the  $\tau$ -th gray level.  $N_1$  is the total number of pixels in the image  $X_1$ .  $n_z$  is the  $z$ -th of the gray level in the image  $X_2$  started from the  $\tau + 1$ -th gray level to the  $L - 1$ -th gray level.  $N_2$  is the total number of pixels in the image  $X_2$ .

Step 6: Calculate CDF in each region is determined as

$$c_1(y) = \sum_{y=0}^{\tau} p_1(y) \quad (12)$$

$$c_2(z) = \sum_{z=\tau+1}^{L-1} p_2(z) \quad (13)$$

Note  $0 \leq c_1(y) \leq 1$  and  $0 \leq c_2(y) \leq 1$  as well.

Step 7: Calculate the mapping function from step 13 in order to represent the new gray level of the resulting image

$$f(y) = \tau[w_1 + c(y) \cdot (1 - w_1)] \quad (14)$$

$$f(z) = (\tau + 1) + [X_{L-1} \cdot w_2 - (\tau + 1)]c_2(z) \quad (15)$$

Step 8: represent the new gray level of the resulting image from step 7 describes as

$$Y = Y_1 \cup Y_2 \quad (16)$$

$$Y = Y_1(f(y)) \cup Y_2(f(z)) \quad (17)$$

The resulting image is equalized independently whereas  $Y$  is a resulting image.  $Y_1$  is the resulting sub-image of sub-histogram  $X_1$  which is equalized by  $f(y)$  and  $Y_2$  is the resulting sub-image of sub-histogram  $X_2$ .

**Table 1** Absolute Mean Brightness Error (AMBE) comparison with traditional methods and our proposed method (WBHE)

Image	GHE	BBHE	WBHE
Airport	45.96031	17.21593	15.43191
Car and APCs	13.93107	11.38658	9.63023
Chemical Plant	23.90703	12.16783	2.30620
Girl	17.37088	12.16783	0.06348
Moon Surface	1.25349	1.81264	1.32314
Splash	25.44266	11.56768	0.41983
Tank	21.76308	12.75653	5.79171
Tiffany	82.16163	38.03572	37.81729
Track and APCs	22.63727	11.74712	0.22270
Truck	23.03217	14.91651	7.54406
U2	98.81740	41.47091	25.85977

#### IV. CONCLUSION

Our proposed method is presented to overcome both GHE and BBHE that existing method. Our proposed method is able to preserve the resulting images with high contrast enhancement as shown in Fig.4 and Fig. 5. Moreover it also protects against AMBE simultaneously as shown in Table 1.

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#### REFERENCES

- [1] Se-Hwan Yun, Jin Heon Kim and Suki Kim, Contrast Enhancement using a Weighted Histogram Equalization, *2011 IEEE International Conference on Consumer Electronics (ICCE)*, 2011, pp. 203 – 204.
- [2] Yeong-Taeg Kim, Contrast Enhancement Using Brightness Preserving Bi-Histogram Equalization, *IEEE Transactions on Consumer Electronics*, Vol. 43, No. 1 February 1997, pp. 1 – 8
- [3] R. C. Gonzalez and P. Wints, *Digital Image Processing*, 2<sup>nd</sup> Edition, Addison-Wesley Publishing Co., Reading, Massachusetts, 1987.
- [4] OGE Marques, *Practical Image and Video Processing Using MATLAB*, A John Wiley and Sons. Inc. Publication, 2011.
- [5] Soong-Der Chen, Ramli and A.R., Minimum Mean Brightness Error Bi-Histogram Equalization in Contrast Enhancement, *2003 IEEE Transactions on Consumer Electronics*, Vol. 49, Issue 4, pp. 1310 – 1319.

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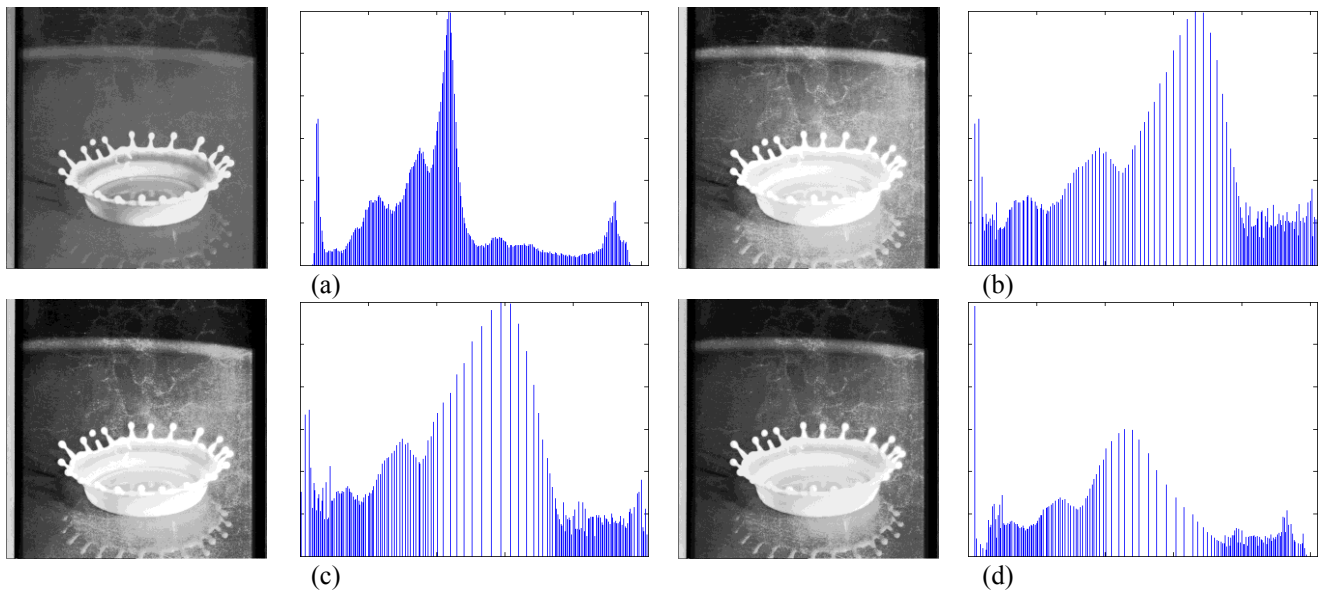


Fig. 4 Splash Images and its corresponding histograms comparison

(a) Original image (b) resulting image of (a) using GHE  
(c) resulting image of (a) using BBHE (d) resulting image of (a) using WBHE and its histogram

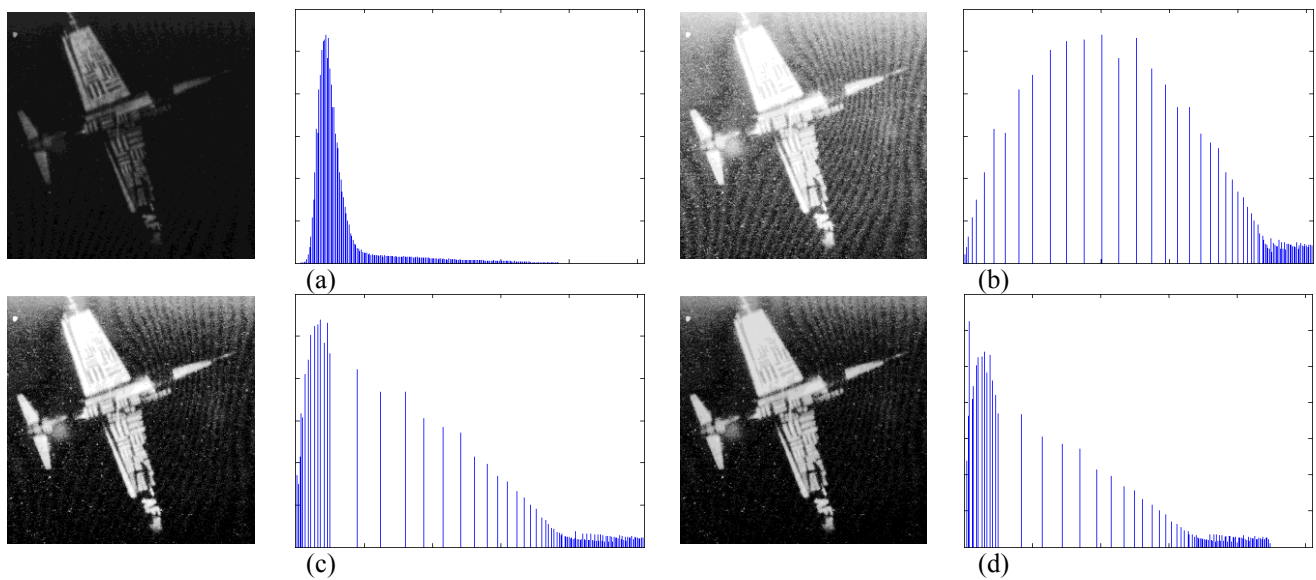


Fig. 5 U2 Images and corresponding histograms comparison

(a) Original image and its corresponding histogram (b) resulting image of (a) using GHE  
(c) resulting image of (a) using BBHE (d) resulting image of (a) using WBHE