Image Enhancement through Contrast Improvement in ROI using Local PGI Model

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ABSTRACT

Images are very powerful tools to provide information to the viewers in every field i.e. medical images for doctors, forensic images for police investigation, text images for readers etc. In the process of image acquisition, image clarity is affected by lighting, weather, distance, or equipment used for image capture. Sometimes quality of the image may be corrupted differently in various regions of an image. As contrast is one of the assessment factors for determining an image quality, it is necessary to develop a better and faster algorithm for contrast improvement in regions of interest. This paper proposes a method for image enhancement through contrast improvement in regions of interest using a Local Parameterized Gradient Intercept (LPGI) Model in spatial domain. The proposed method provides good results subjectively as well as objectively for both gray scale and true color images. The proposed method is useful for interactive image processing applications as it has a family of possible transformations for various enhancement levels in different regions of interest.

Keywords

Global Histogram Equalization (GHE), Local Histogram Equalization (LHE), Global Adaptive Equalization (GAE), Local Adaptive Equalization (LAE), and Global Parameterized Gradient Intercept (GPGI), Local Parameterized Gradient Intercept (LPGI), Region Of Interest (ROI).

1. INTRODUCTION

Image enhancement improves digital image quality without knowing the source of degradation and provides visually acceptable images for human viewers and/or automated image processing techniques. We reviewed enhancement techniques for gray scale images in spatial domain [1] and implemented them using MATLAB. These techniques have been successfully extended to true color images [2]. Image enhancement process gives better visual quality either by increasing the contrast or suppressing the noise. Image enhancement can be done through contrast improvement using a linear PGI model [3]. Image enhancement through noise suppression can be done using a nonlinear PAR model [4]. Image enhancement through contrast improvement and noise suppression can be done using a Parameterized Hybrid model [5]. These three global models: linear, nonlinear, and hybrid work well when entire image is corrupted uniformly. They fail to enhance when various regions of an image corrupted differently. This paper proposes a method for image enhancement through contrast improvement in ROI using a Local PGI model in spatial domain.

2. PROPOSED METHOD

Let f(x,y) be a digital image of size MxN and g(x,y) be its enhanced image along with respective pixel values $\, r \,$ and $\, s \,$. The transformation for proposed model for $i=1,2,3,\ldots n \,$ is given by

$$g_i(x, y) = G_i * f_i(x, y) + I_i$$
 (1)

$$\Rightarrow s_i = G_i * r_{i+1} I_i$$
 (2)

where n is the number of regions of interest, G_i and I_i are the Gradient and Interception of ith region of interest. G_i and I_i values can be zero, positive, or negative. When G_i and/or I_i values are varied for improving the image contrast, above transformation becomes simple linear or nonlinear but not exponential or logarithmic as in traditional point processing methods [6][7]. Proposed method needs simple operations and does not require PDF calculations for finding histogram of an image as in GHE and GAE methods [8].

The proposed method is given the name 'Local Parameterized Gradient Intercept (LPGI) model', as a family of possible transformations can be obtained for achieving effective image enhancement for region of interest.

3. LPGI ALGORITHM

The following are the steps involved in image enhancement through contrast improvement in ROI using LPGI algorithm for gray scale and true color images.

Gray scale image:

- 1. Read input image i(x,y)
- 2. Corrupt the image i(x,y) to get f(x,y)
- 3. Divide f(x,y) into two sub images based on ROI: $f_1(x,y)$ and $f_2(x,y)$
- 4. For $f_1(x,y)$, select appropriate G_1 and I_1 .
- 5. Multiply $f_1(x,y)$ with G_1 and add with I_1 to get $g_1(x,y)$.
- 6. If $g_1(x,y)$ is not good in contrast change G_1 and/or I_1 , then go to step5
- 7. For $f_2(x,y)$, select appropriate G_2 and I_2 .
- 8. Multiply $f_2(x,y)$ with G_2 and then add I_2 to get $g_2(x,y)$.
- If g₂(x,y) is not good in contrast change G₂ and/or I₂, then go to step 8.
- 10. Get enhanced output image g(x,y) from $g_1(x,y)$ and $g_2(x,y)$.

True Color image:

- 1. Read input image i(x,y).
- 2. Corrupt the image i(x,y) to get f(x,y)
- 3. Divide f(x,y) into two sub images based on ROI: $f_1(x,y)$ and $f_2(x,y)$

- 4. For $f_1(x,y)$, select values of G_1 and I_1 .
- 5. Extract r,g,b components from $f_1(x,y)$
- 6. Multiply $r_1(x,y)$, $g_1(x,y)$, and $b_1(x,y)$, with G_1 and add with I_1 to get $h_1(x,y)$.
- If h₁(x,y) is not good in contrast change G₁ and/or I₁, then go to step 6.
- 8. For $f_2(x,y)$, select values of G_2 and I_2 .
- 9. Extract r,g,b components from $f_2(x,y)$
- 10. Multiply $r_2(x,y)$, $g_2(x,y)$, and $b_2(x,y)$, with G_2 and add with I_2 to get $h_2(x,y)$.
- 11. If $h_2(x,y)$ is not good in contrast change G_2 and/or I_2 , then go to step 10.
- 12. Get enhanced output image h(x,y) from $h_1(x,y)$ and $h_2(x,y)$.

4. RESULTS

The LPGI model performance can be compared to that of GHE, LHE, GAE, LAE, and GPGI methods. Consider two gray scale images of file TIF type: Lena(512x512) and Man(256x256) and then two true color images of file JPG type: Pepper(384x512) and Thajl(256x256). These images are shown in figures Fig. (1-4) have been corrupted differently in two regions. Corrupted images are then enhanced by Global methods: GHE, GAE, and GPGI. Let the corrupted image can be segmented into two regions ROI₁ and ROI₂. They are either left half and right half or top half and bottom half. Segmented images are then enhanced by local methods: LHE, LAE, and LPGI. Contrast improvement can be judged by visual inspection of the resultant images along with their histograms and also by evaluating the mean square error for each method [9-15].

$$MSE = \frac{1}{MN} \sum_{i=1}^{M} \sum_{j=1}^{N} [f(i,j) - g(i,j)]^{2}$$
 (3)

Subjective results for gray scale images are shown in figures (Fig. 5-20) and for true color images in figures (Fig. 21-36). Objective results for gray scale images are given in table 1 and for true color images in tables 2.

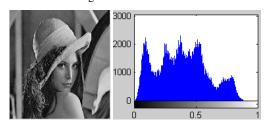


Fig 1: Lena image and its histogram

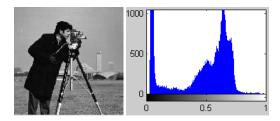


Fig 2: Man image and its histogram

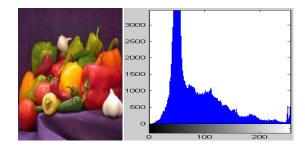


Fig 3: Pepper image and its histogram

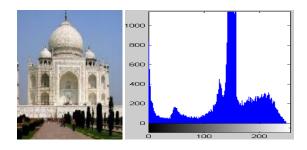


Fig 4: Thaj image and its histogram

Table 1. MSE for gray scale images

Image	Lena	Man
Original	1.1686	1.0378
GHE	0.1187	0.1620
LHE	0.1188	0.0999
GAE	0.1245	0.1741
LAE	0.1257	0.1753
GPGI1	0.0600	23.526
GPGI2	8.0496	0.1512
LPGI	0.0001	0.0001

Table 2. MSE for true color images

Image	Pepper			
	MSE _R	MSE_G	MSE _B	
Original	0.0938	0.0987	0.0849	
GHE	0.0680	0.0794	0.0737	
LHE	0.0577	0.0821	0.0787	
GAE	0.0787	0.0907	0.0802	
LAE	0.0799	0.0912	0.0803	
GPGI1	0.0570	0.0993	0.0953	
GPGI2	0.0471	0.0191	0.0127	
LPGI	1.94e-5	2.04e-5	2.03e-5	
Image	Thaj			
	MSE _R	MSE _G	MSE _B	
Original	0.0730	0.0839	0.1019	
GHE	0.0464	0.0561	0.0697	
LHE	0.0341	0.0402	0.0443	
GAE	0.0631	0.0737	0.0898	
LAE	0.0633	0.0737	0.0902	
GPGI1	0.0422	0.0455	0.0524	
GPGI2	0.0536	0.0536	0.0868	
LPGI	2.04e-5	2.05e-5	2.05e-5	

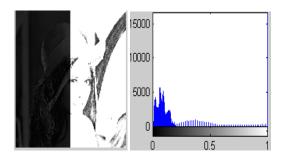


Fig 5: Corrupted image and its histogram

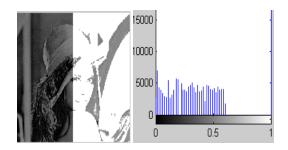


Fig 6: GHE image and its histogram

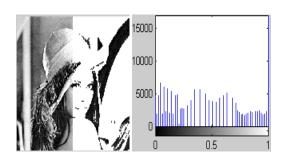


Fig 7: LHE image and its histogram

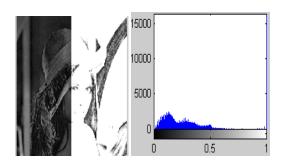


Fig 8: GAE image and its histogram

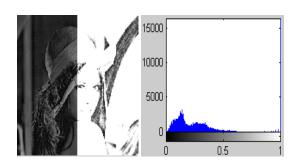


Fig 9: LAE image and its histogram

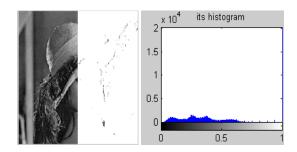


Fig 10: GPGI1 image and its histogram

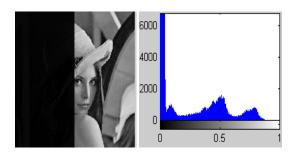


Fig 11: GPGI2 image and its histogram

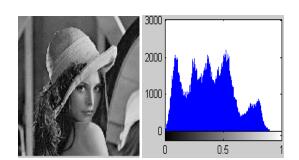


Fig 12: LPGI image and its histogram

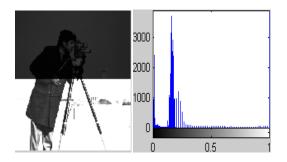


Fig 13: Corrupted image and its histogram

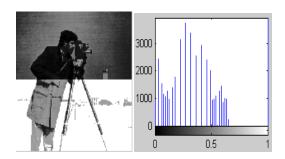


Fig 14: GHE image and its histogram

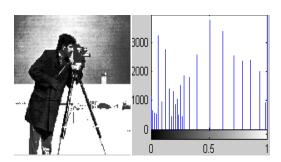


Fig 15: LHE image and its histogram

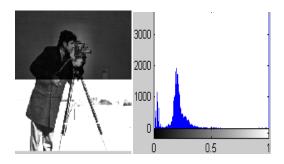


Fig 16: GAE image and its histogram

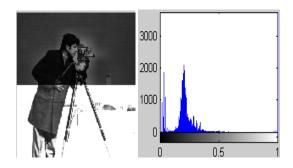


Fig 17: LAE image and its histogram

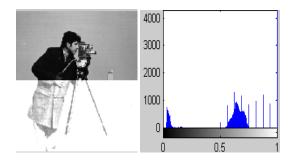


Fig 18: GPGI1 image and its histogram

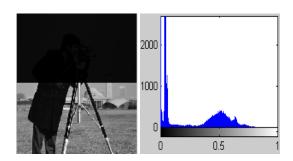


Fig 19: GPGI2 image and its histogram

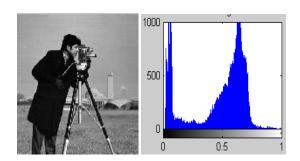


Fig 20: LPGI image and its histogram

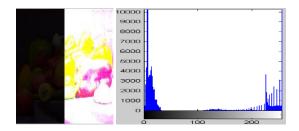


Fig 21: Corrupted image and its histogram

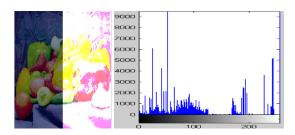


Fig 22: GHE image and its histogram

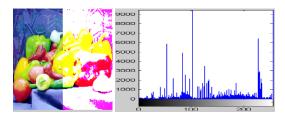


Fig 23: LHE image and its histogram

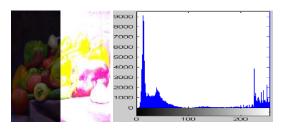


Fig 24: GAE image and its histogram

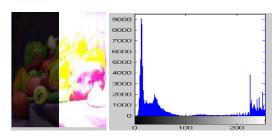


Fig 25: LAE image and its histogram

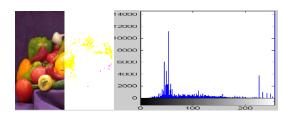


Fig 26: GPGI1 image and its histogram

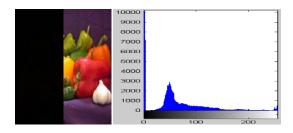


Fig 27: GPGI2 image and its histogram

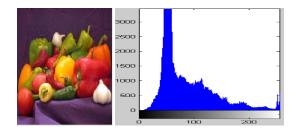


Fig 28: LPGI image and its histogram

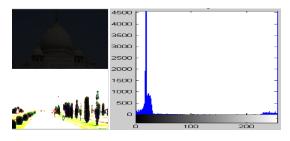


Fig 29: Corrupted image and its histogram

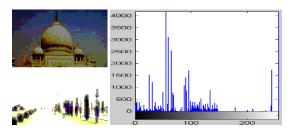


Fig 30: GHE image and its histogram

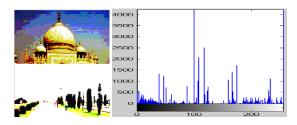


Fig 31: LHE image and its histogram

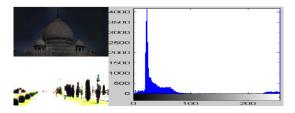


Fig. 32: GAE image and its histogram

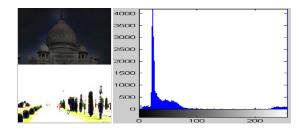


Fig 33: LAE image and its histogram

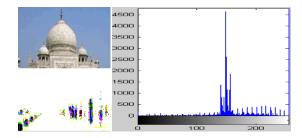


Fig 34: GPGI1 image and its histogram

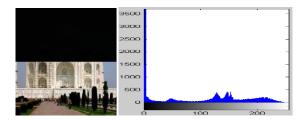


Fig 35: GPGI2 image and its histogram

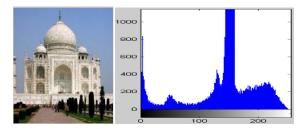


Fig 36: LPGI image and its histogram

5. DISCUSSIONS

Visual inspection of subjective results indicates that, GHE, LHE, GAE, LAE methods fail to enhance images that are corrupted differently in two regions. GPGI method enhances only one region where as proposed LPGI model works very well by enhancing both regions of gray scale and true color images. Visual inspection of objective results shows that LPGI method has smaller mean square error for the gray scale images and also for the R, G, and B components of true color images. Simulation results for an image are restricted to two regions only. Proposed algorithm can be extended to any number of regions. The proposed model can be applied to enhance the images that are segmented in rectangular or square shape only. This limitation can be overcome by slight change in LPGI algorithm for selecting spatial coordinates to enhance the any shape of ROI.

6. CONCLUSIONS

Image enhancement through contrast improvement in ROI using LPGI has been successfully implemented for both gray

scale and true color images in spatial domain. LPGI model works well for a gray scale images and results are much more pronounced for true color image by preserving maximum color details. Choice of G and I depend on the type of image. LPGI method can be used as a tool for Photo editing software like Photoshop or any existing image processing software by providing two sliding bars for selecting spatial coordinates x and y of ROI and then two more sliding bars for G and I for image enhancement. Sometimes an image can be corrupted by different amounts of noise intensities in its various regions. The future scope will be the development of local parameterized model for achieving image enhancement through noise suppression in ROI from noisy images.

7. REFERENCES

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