

# MORPHOLOGICAL BACKGROUND DETECTION AND CONTRAST ENHANCEMENT OF IMAGES IN MATLAB

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## ABSTRACT

This paper deals with enhancement of images with poor contrast and detection of background and proposes new method for poor lighting contrast enhancement of various images based on the Weber's law. Background of various images is identified by using the contrast enhancement transformations. The contrast image transformation can be defined by two operators opening and closing. The first operator employs information from block analysis, while the second transformation utilizes the opening by reconstruction. Opening by reconstruction is used to define the multi background notion. The complete image processing is done using MATLAB simulation model. This paper also explain Morphological transformation and Weber's law. Image background approximation is applied to the background by means of block analysis in conjunction with transformations that enhance images with poor lighting. The multi background notion is introduced by means of the opening by reconstruction which shows a comparison among several techniques to improve contrast in images.

**Keywords:** Morphological, Weber's law, Contrast Enhancement

## 1. INTRODUCTION

It is very necessary to enhance the contrast of which images before further processing or analysis can be conducted. Image enhancement is a challenging research area in image processing. Its principal objective is to process an image so that the result is more suitable than the original image for a specific application. There are several techniques to enhance an image. Image enhancement is a technique that increases the visual contrast of an image in a designated intensity range or ranges.

The background in poor lighting of various grey level images can be identified by the use of morphological operators. Then image enhancement has been carried out by the application based on Weber's law. Thereafter erosion , dilation and opening by reconstruction method is followed [1].

## 2. CONTRAST ENHANCEMENT

Contrast is the opposition or dissimilarity of things that are compared ,extent to which adjacent areas on the image differ in brightness. Contrast is act of distinguishing by comparing differences. Contrast refers to how bright the highlights are while brightness refers to how bright the shadows are. For example, if you had a dark rectangle and light rectangle in the image , the brightness setting will affect the dark rectangle and contrast setting will affect the light rectangle. If both brightness and contrast are set to minimum, the screen will be pure black[2].

The image enhancement includes the improvement of the visibility and perceptibility of the various regions, in which an image can be partitioned and defect ability of the image features inside the regions. Image enhancement is usually followed by detection of features such as edges, peaks, and other geometric features which is of paramount importance in low-level vision. One can see that the enhanced image is a shaper image than the original. Particularly interesting is the fact that textural information has been rendered visible in the enhanced image[3][4].

### **3. MORPHOLOGICAL TRANSFORMATION AND WEBER'S LAW**

#### **3.1. Morphology**

Morphology is a technique of image processing based on shapes. The value of each pixel in the output image is based on a comparison of the corresponding pixel in the input image with its neighbors. By choosing the size and shape of the neighborhood, morphological operation can be constructed that is sensitive to specific shapes in the input image[5].

Mathematical morphology is a set-theoretical approach to multi-dimensional digital signal or image analysis, based on shape. It is a theory and technique for the analysis and processing of geometrical structures, based on set theory, lattice theory, topology, and random functions. It is most commonly applied to digital images, but it can be employed as well on graphs, surface meshes, solids, and many other spatial structures[6].

It is also the foundation of morphological image processing, which consists of a set of operators that transform images according to the above characterizations. Mathematical morphology was originally developed for binary images, and was later extended to scale functions and images. The subsequent generalization to complete lattices is widely accepted today as MM's theoretical foundation.

#### **3.2. Morphological transformations**

Basically morphological transformations such as erosion, dilation, opening and closing are used to detect the background.

##### **3.2.1. Erosion**

Erosion is one of the two basic operators in the area of mathematical morphology, the other being dilation. It is typically applied to binary images, but there are versions that work on gray scale images. The erosion operator takes two pieces of data as inputs. The first is the image which is to be eroded. The second is a (usually small) set of coordinate points known as structuring elements (also known as a kernel). It is this structuring element that determines the precise effect of the erosion on the input image. In erosion, every object pixel that is touching a background pixel is changed into a background pixel.

Gray scale erosion with a flat disk shaped structuring element will generally darken the image. Bright regions surrounded by dark regions shrink in size, and dark regions surrounded by bright regions grow in size. Small bright spots in images will disappear as they are eroded away down to the surrounding intensity value, and small dark spots will become larger spots. The effect is mostly marked at places in the image where the intensity changes rapidly, and regions of fairly uniform intensity will be left more or less unchanged except at their edges[7].

##### **3.2. 2. Dilation**

Dilation adds pixels to the boundaries of objects in an image. In dilation, every background pixel that is touching an object pixel is changed into an object pixel. Note how the function applies the rule to the input pixel's

neighborhood and uses the highest value of all the pixels in the neighborhood as the value of the corresponding pixel in the output image[8]. The dilation operator takes two pieces of data as inputs. The first is the image which is to be dilated. The second is a (usually small) set of coordinate points known as a structuring element (also known as a kernel ). It is this structuring element that determines the precise effect of the dilation on the input image. To compute the dilation of a binary image by structuring element , Each of the background pixels in the input image in turn is considered. For each background pixel, the structuring element coinciding with the input pixel position is superimposed.

### 3.3. Weber's law

Weber's law states that, it is the ratio of the difference in maximum to minimum luminance value to the minimum luminance value and it is denoted by C[9] .

$$C = \frac{L_{\max} - L_{\min}}{L_{\min}} \quad (1)$$

$$\text{If } L = L_{\min} \quad \Delta L = L_{\max} - L_{\min}$$

can be rewritten as

$$C = \frac{\Delta L}{L}$$

This indicates  $\Delta(\log L)$  is proportional to C; therefore, Weber's law can be expressed as

$$C = k \log L + b \quad L > 0 \quad (2)$$

Where, k and b are constants, b being the background.

## 4. IMAGE BACKGROUND DETECTION BY BLOCK ANALYSIS

In this analysis, first of all an image will be read as input image and divided into several blocks and from each block ,the background will be determined and after applying the Weber's law , an enhanced image will be obtained [10].

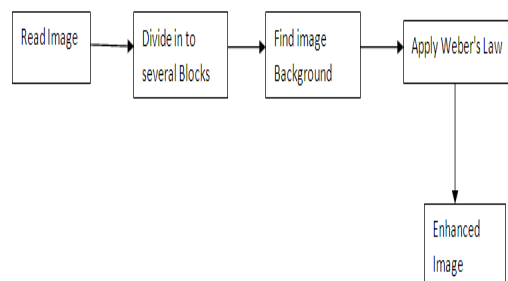


Fig.1. Block diagram of Background Detection by Block Analysis

Let us consider an image which is to be enhanced. The image is divided into  $n$  blocks of size. Each block is a sub image of the original image. As the image is made up of number of pixels each block consists of number of pixels. Find the maximum and minimum intensity values of pixels of each block [11].

For each analyzed block, maximum ( $M_i$ ) and minimum ( $m_i$ ) values are used to determine the background criteria  $T_i$  in the following way:

$$\tau_i = \frac{m_i + M_i}{2} \quad \forall i = 1, 2, \dots, n \quad (3)$$

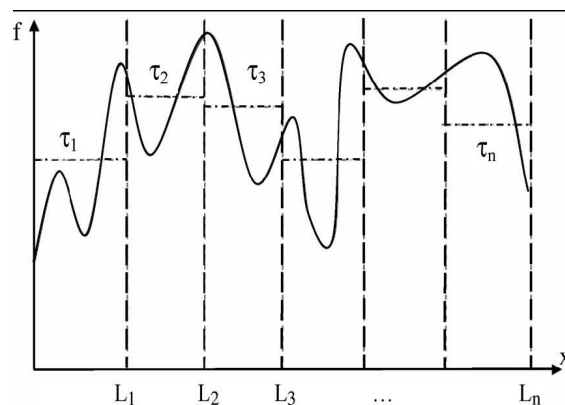


Fig 2. Background criteria obtained by block analysis.

In the 1-D case, as illustrated in Figure 2, the following expression is obtained:

Once  $T_i$  is calculated, this value is used to select the background parameter associated with the analyzed block. As follows, an expression to enhance the contrast is proposed:

$$\Gamma \tau_i(f) = \begin{cases} k_i \log(f + 1) + M_i, & f \leq \tau_i \\ k_i \log(f + 1) + m_i, & \text{otherwise} \end{cases} \quad (4)$$

Note that the background parameter depends on the  $T_i$  value. If  $f \leq T_i$  (dark region), the background parameter takes the value of the maximum intensity ( $M_i$ ) within the analyzed block, and the minimum intensity ( $m_i$ ) value otherwise. Also, the unit was added to the logarithm function in above equation to avoid indetermination [12]. On the other hand, since grey level images are used in this work, the constant  $K_i$  in above equation is obtained as follows:

$$k_i = \frac{255 - m_i}{\log(256)} \quad \forall i = 1, 2, \dots, n \quad (5)$$

$$m_i = \begin{cases} m_i, & f > \tau_i \\ M_i, & f \leq \tau_i \end{cases} \quad (6)$$

On the other hand,  $M_i$  and  $m_i$  values are used as background parameters to improve the contrast depending on the  $T_i$  value, due to the background is different for clear and dark regions. Now an image is formed by applying

the equation number 6. Now consider a pixel in this image and the corresponding pixel in original image. Combine them using Weber's law which can be stated as follows. Thus, an enhanced image is formed[13][14].

### 5. IMAGE BACKGROUND DETERMINATION USING OPENING BY RECONSTRUCTION

On the other hand, given that, maximum and minimum values are analyzed for each block, an extension using morphological operators is presented as follows.

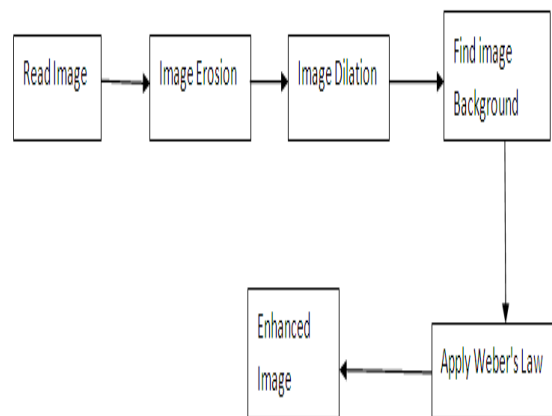


Fig.3. Block diagram of Background Detection by Erosion & Dilation

In a morphological operation, the value of each pixel in the output image is based on a comparison of the corresponding pixel in the input image with its neighbors. By choosing the size and shape of the neighborhood, a morphological operation can be constructed that is sensitive to specific shapes in the input image. Dilation and erosion are two fundamental morphological operations. Dilation adds pixels to the boundaries of objects in an image, while erosion removes pixels on object boundaries. The number of pixels added or removed from the objects in an image depends on the size and shape of the structuring element used to process the image. In the morphological dilation and erosion operations, the state of any given pixel in the output image is determined by applying a rule to the corresponding pixel and its neighbors in the input image[15] [16].

In erosion, every object pixel that is touching a background pixel is changed into a background pixel. Dilation adds pixels to the boundaries of objects in an image. In dilation, every background pixel that is touching an object pixel is changed into an object pixel. Note how the function applies the rule to the input pixel's neighborhood and uses the highest value of all the pixels in the neighborhood as the value of the corresponding pixel in the output image.

Let  $I_{\max}(x)$  and  $I_{\min}(x)$  be the maximum and minimum intensity values taken from one set of pixels contained in a window (B) of elemental size (3 X 3 elements), x belongs to D. Notice that the window corresponds to the structuring element. A new expression can be derived as shown:

$$\tau(x) = \frac{I_{\min}(x) + I_{\max}(x)}{2} \quad (7)$$

Where  $I_{\max}(x)$  and  $I_{\min}(x)$  values correspond to the morphological dilation and erosion defined by the order-statistical filters. Thus, the above expression is expressed as

$$\tau(x) = \frac{\epsilon_{\mu}(f)(x) + \delta_{\mu}(f)(x)}{2} \quad (8)$$

Finally the proposed transformation is expressed as

$$\Gamma_{\tau(x)}(f) = \begin{cases} k_{\tau(x)} \log(f+1) + \delta_{\mu}(f)(x), & f \leq \tau(x) \\ k_{\tau(x)} \log(f+1) + \epsilon_{\mu}(f)(x), & \text{otherwise} \end{cases}$$

and  $k_{\tau(x)} = \frac{255 - \tau(x)}{\log(256)}$  (9)

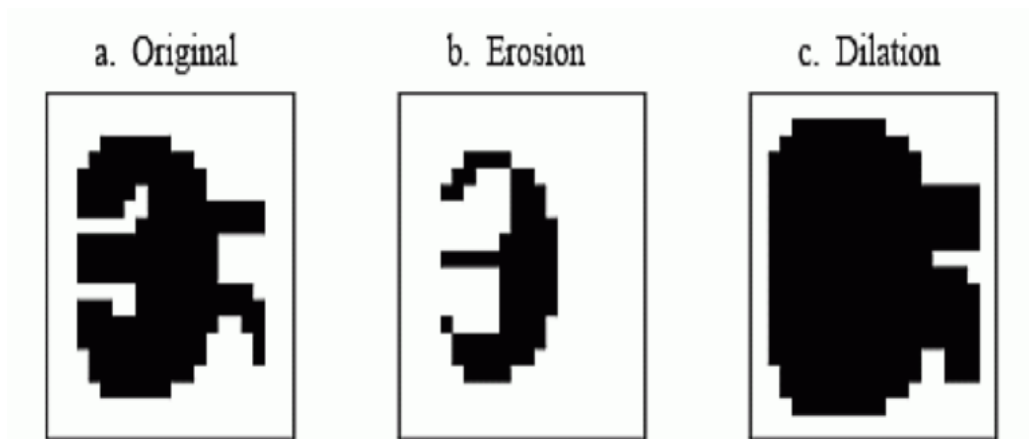


Fig. 4. a) Original Image b) Erosion c) Dilation

## 6. HISTOGRAM EQUALIZATION

Histogram equalization is one of the well known image enhancement technique. It became a popular technique for contrast enhancement because this method is simple and effective. In the latter case, preserving the input brightness of the image is required to avoid the generation of non-existing artifacts in the output image. Although these methods preserve the input brightness on the output image with a significant contrast enhancement, they may produce images which do not look as natural as the input ones. The basic idea of histogram equalization method is to

re-map the gray levels of an image. Histogram equalization tends to introduce some annoying artifacts and unnatural enhancement.

This technique is widely used to improve images with poor lighting. The histogram equalization technique consists in reordering the grey level intensities within the image to obtain a uniformly distributed histogram[17][18].

## 7. EXPERIMENT RESULT ANALYSIS AND DISCUSSION

The proposed method for enhancing the various poor lighting images have been implemented using MATLAB tool. The result of each method are analysed by using statistical parameter. The detailed descriptions are given below.

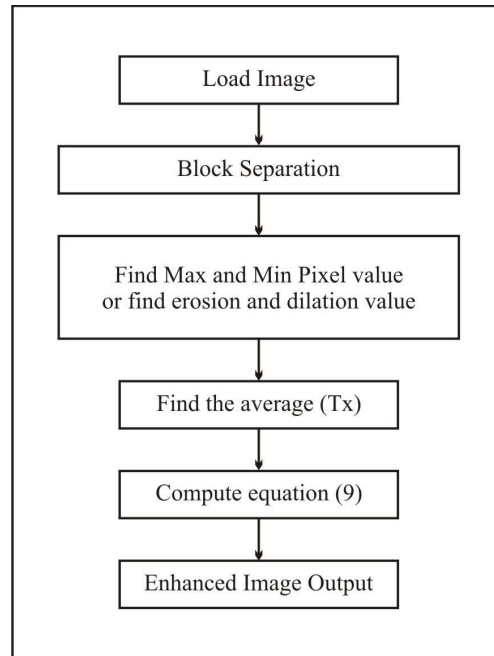


Fig. 5. The architecture of the proposed system

The logical descriptions are given below.

- 1) Preprocess to get data from image (i.e. color data retrieve of image )
- 2) Pixelgrabber retrieve image data in 1 D array.
- 3) R G B component are now separated
- 4) Converts RGB into Black & White . This data is linear data ( 1 D data)
- 5) For image processing ,converts it into 2D
- 6) For each pixel , gets (Known) min & max value of surrounding pixel.  
Meanv= (maxv+minv)/2.0
- 7) For each pixel , value mv is greater or min
- 8) If max  $K=(255-maxv)$  , If less  $K=(255-minv)$

## 8. CONCLUSION

This paper presents the method to detect the image background and to enhance the contrast in grey level images with poor lighting. First, a methodology was introduced to compute an approximation to the background using blocks analysis. This proposal was subsequently extended using mathematical morphology operators. However, a difficulty was detected when the morphological erosion and dilation were employed, therefore, a new methodology to detect the image background was propounded, that is based on the use of morphological connected transformations.

Also, morphological contrast enhancement transformations were introduced. Such operators are based on Weber's law notion. These contrast transformations are characterized by the normalization of grey level intensities, avoiding abrupt changes in illumination. The performances of the proposals provided in this work were illustrated by means of several examples throughout the paper. Also, the operators performance employed in this paper were compared with others given in the literature. Finally contrast enhancement transformations studied in this paper is that they can only be used satisfactorily in images with poor lighting .

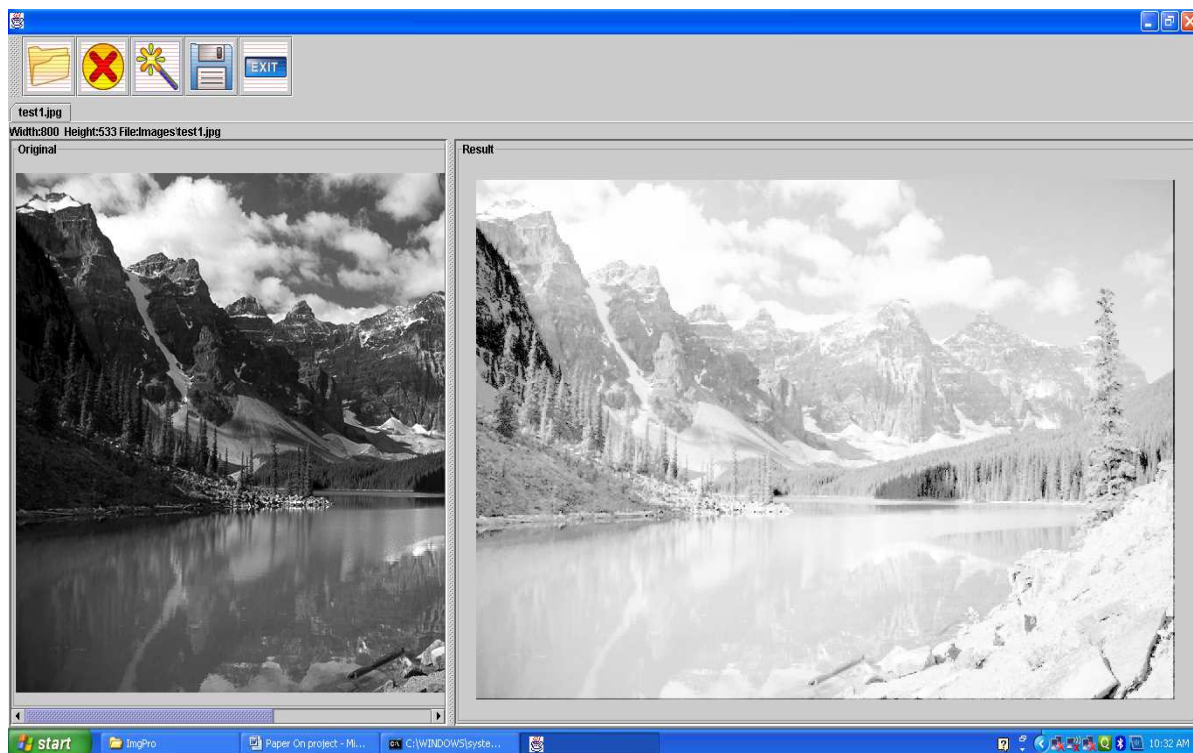


Fig. 6. a) Original Image

b) Output Image



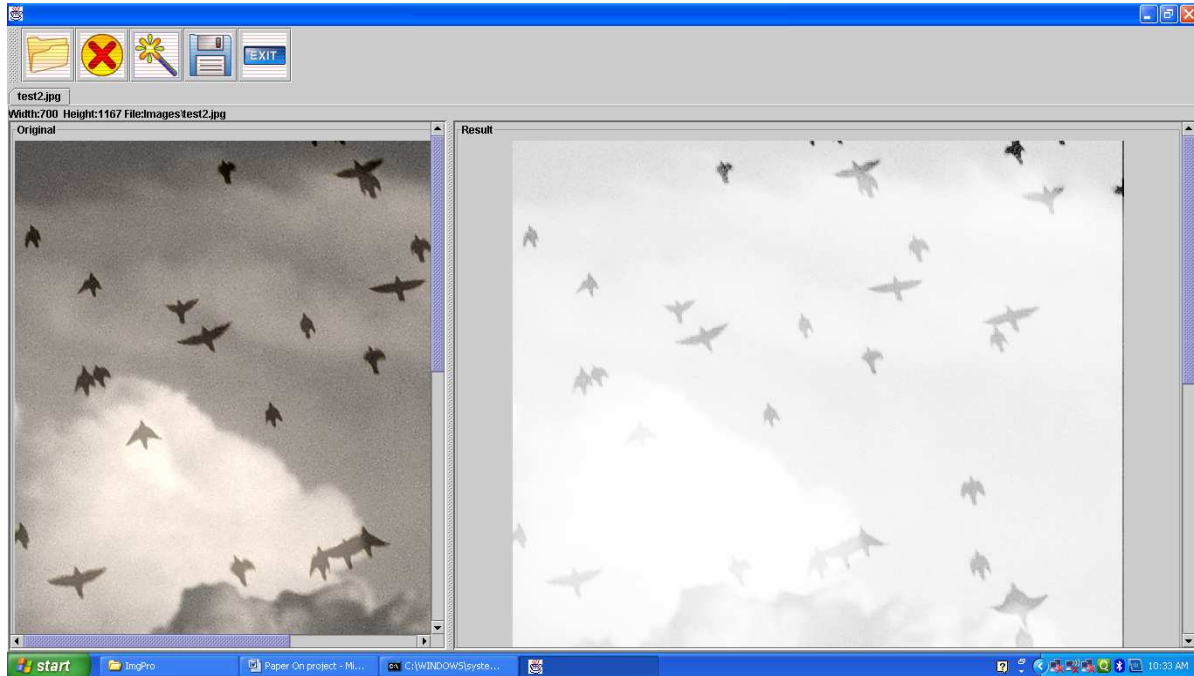


Fig. 7. a) Original Image                      b) Output Image

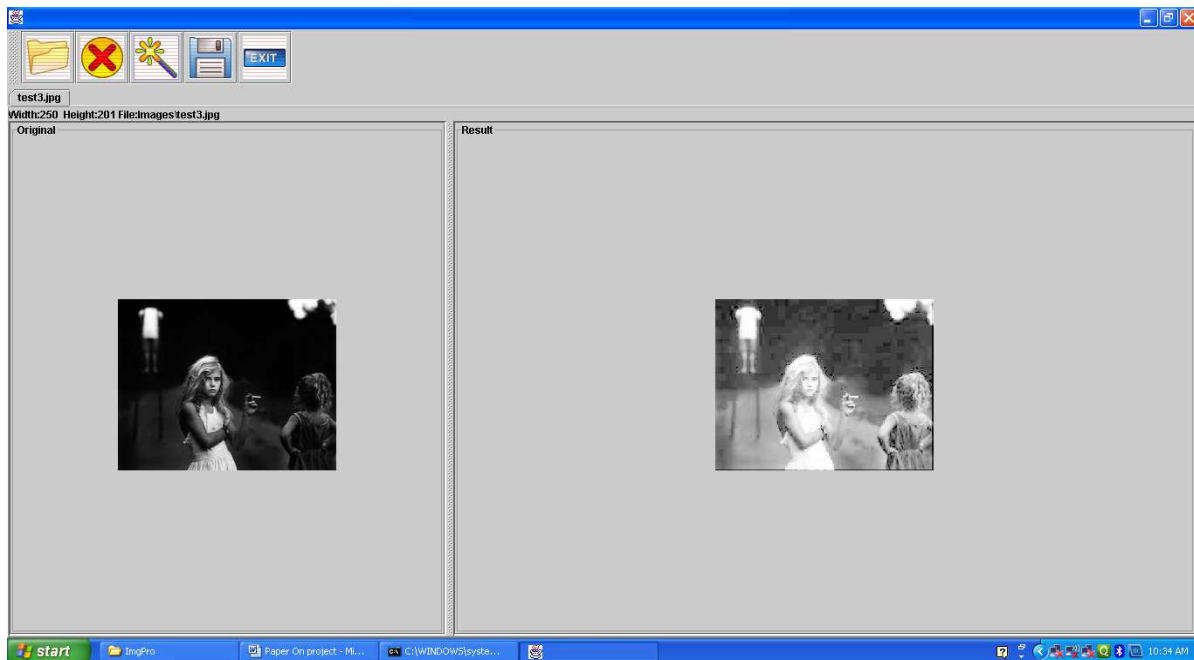


Fig. 8. a) Original Image                      b) Output Image

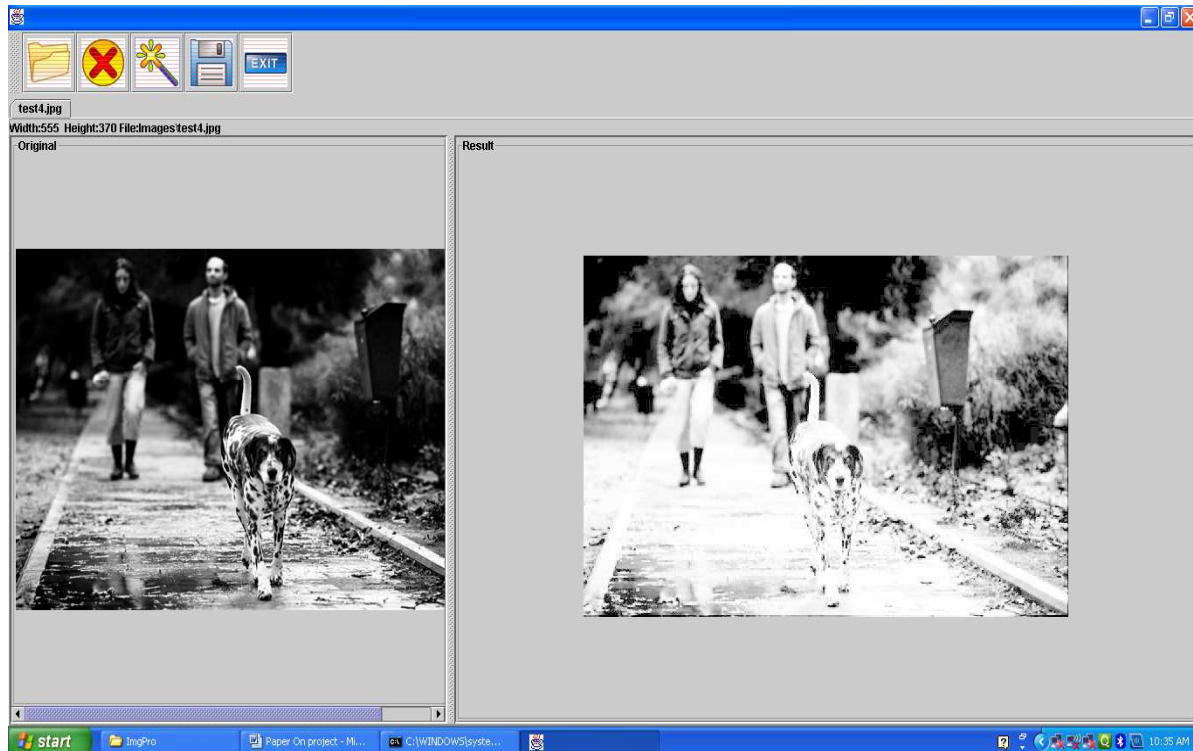


Fig. 9. a) Original Image

b) Output Image

### References :

- [1] S. and R. Manavalan , "Analysis of Background Detection and Contrast Enhancement of MRI Images," International Journal of Computer Applications (0975 – 8887) Volume 36–No.12, December 2011.
- [2] K.R.Ananth and Dr. S.Pannerselvam "Automated CAD for Nodule Detection for Magnetic Resonance Image Contrast Enhancement," 2011
- [3] I.R.Terol-Villalobos, "Morphological connected contrast mappings based on top-hat criteria: A multi scale contrast approach," *Opt. Eng.*, vol. 43, no. 7, pp. 1577–1595, 2004.
- [4] K.Sreedhar and B.Panlal "Enhancement of Images using Morphological Transformation," International Journal of Computer Science & Information Technology (IICSIT) Vol 4, No 1, Feb 2012 .
- [5]. I.R. Terol-Villalobos, "Morphological image enhancement and segmentation with analysis," P. W. Hawkes, Ed. New York: Academic, 2005, pp. 207–273.
- [6]. I. R. Terol-Villalobos, "A multiscale contrast approach on Morphological connected contrast Mappings," *Opt. Eng.*, vol. 43, no. 7, pp. 1577–1595, 2009.
- [7]. J. Kasperek, "Real time morphological image contrast enhancement in FPGA," in LNCS, New York: Springer, 2008.
- [8] Jianhong Shen "Weber's Law and Weberized TV Restoration," School of Mathematics University of Minnesota Minneapolis, MN 55455, USA.□
- [9] A.I. BEENA and B. ARTHI , "Contrast enhancement in gray-scale images using background approximation by blocks and morphological operations," Oriental Journal of Computer Science & Technology Vol. 3(1), 69-73 (2010).
- [10]. A.Majumder and S. Irani, "Perception-based contrast enhancement of images," *ACM Trans. Appl. Percpt.*, 4(3): (2007).
- [11]. Angélica R. Jiménez-Sánchez, Jorge D.Mendiola-Santibañez, "Morphological Background Detection and enhancement of Images with Poor Lighting," *IEEE Trans. Image Process.* 18(3), pp. 613-623 (2009) .
- [12] Tinu Alexander John , G. Muthupandi M.E " Background Detection of Image using Approximation by Block and Opening by Reconstruction Transformation," International Conference on Emerging Technology Trends (ICETT) 2011 Proceedings published by International Journal of Computer Applications® (IJCA) .

- [13] Soille, P., "Morphological image analysis: principles and applications," Springer Verlag, Berlin (2003).
- [14] Z. Liu, C. Zhang, and Z. Zhang, "Learning-based perceptual image quality improvement for video conferencing," presented at the IEEE Int. Conf. Multimedia and Expo (ICME), Beijing, China, Jul. 2007.
- [15] J. Serra and P. Salembier, "Connected operators and pyramids," presented at the SPIE. Image Algebra and Mathematical Morphology, San Diego, CA, Jul. 1993.
- [16]. A. Toet, "Multiscale contrast enhancement with applications to image fusion," Opt. Eng., vol. 31, no. 5, 1992.
- [17]. S. Mukhopadhyay and B. Chanda, "A multiscale morphological approach to local contrast enhancement," Signal Process. vol. 80, no. 4, pp. 685-696, 2000.
- [18]. Erik R. Urbach and Michael H. F. Wilkinson "Efficient 2-D Grayscale Morphological Transformations With Arbitrary Flat Structuring Elements," IEEE TRANSACTIONS ON Image Processing, VOL. 17, NO. 1, January 2008,