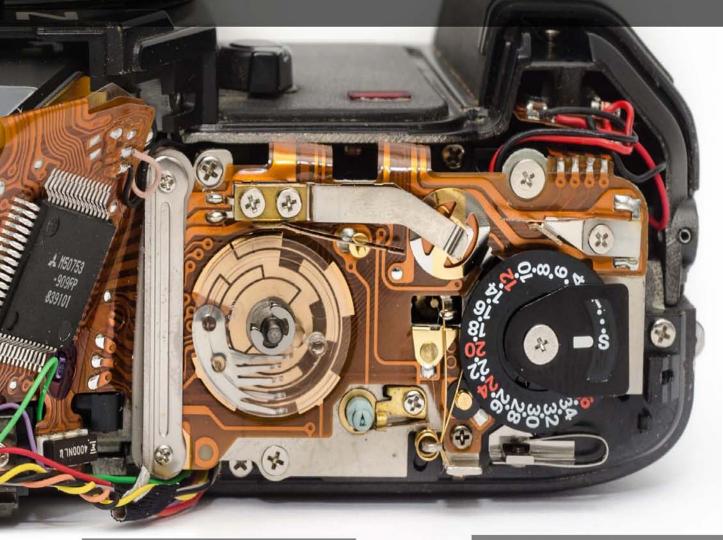


Global Journal

OF COMPUTER SCIENCE AND TECHNOLOGY: F

Graphics & Vision



An Ambigramic Image File Format

Mammogram Image Segmentation

Highlights

The Impact of Different Image

Computer Generation of Fractals

Discovering Thoughts, Inventing Future

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An Ambigramic Image File Format

By Wenyi Cui, Kohei Inoue & Kenji Hara

Kyushu University

Abstract- We propose an image file format that can be read in two ways, where two images are recorded in a single file as a bit sequence, and the forward reading the bit sequence makes one of the two images visible, or the backward reading makes another image visible. Such a way of looking at a binary data in two ways resembles that of an ambigram, which is a piece of calligraphy that can be read in two ways by rotating it or introducing other perspectives. The proposed ambigramic image file format is compared with the graphics interchange format (GIF) experimentally, and the results show the better quality of the ambigramic images than that of GIF images.

Keywords: ambigram, image file format, error diffusion, integral image-based signal-to-noise ratio (ISNR).

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An Ambigramic Image File Format

Wenyi Cui^α, Kohei Inoue^σ & Kenji Hara^ρ

Abstract- We propose an image file format that can be read in two ways, where two images are recorded in a single file as a bit sequence, and the forward reading the bit sequence makes one of the two images visible, or the backward reading makes another image visible. Such a way of looking at a binary data in two ways resembles that of an ambigram, which is a piece of calligraphy that can be read in two ways by rotating it or introducing other perspectives. The proposed ambigramic image file format is compared with the graphics interchange format (GIF) experimentally, and the results show the better quality of the ambigramic images than that of GIF images.

Keywords: ambigram, image file format, error diffusion, integral image-based signal-to-noise ratio (ISNR).

Introduction

mbigram is a typographical design that can be read in multiple orientations as shown in Fig. 1, where a word 'ambigram' is written in a rotationally symmetric manner. There are other types of ambigrams such as mirror, perceptual shift and 3D ambigrams [1]. Langdon [2], [3] has made a lot of interesting ambigrams for a long time. Those ambigramic artworks give pleasure to the viewers, and provide great inspiration for artists and engineers.



Figure 1: Example of ambigram

Inspired by such successful ambigrams, in this paper we suggest an application of the idea of ambigram to an ambigramic interpretation of a bit sequence. In other words, we propose a method for describing two images with a bit sequence, which is a digital ambigram that can be seen as one of the two images if the bit sequence is read forward direction, or as another image if the bit sequence is read backward direction. The proposed ambigramic images are compared with the graphics interchange format (GIF) images [4], and the effectiveness of the proposed method is demonstrated in the experiments using natural images, where an image quality measure is used

palette-based images such as GIF images. The rest of this paper is organized as follows: Section 2 describes the proposed algorithm for

for evaluating the quality of halftone images including

generating ambigramic bit sequences. Section 3 shows experimental results. Finally, Section 4 concludes this paper.

Proposed Ambigramic Image Data II. STRUCTURE

In this section, we propose a method for formatting image data into an ambigramic data structure, which is recorded in a binary file. Assume that two images are given as an input data as $A = [a_{ij}]$ and B= $[b_{ij}]$, where a_{ij} and b_{ij} denote color vectors at the pixel position (i; j) in the images A and B, respectively, for i =1; 2; :::; m and j = 1; 2; :::; n, i.e., A and B are the same size. Then we attempt to store A and B in a storage as a single file. Figure 2 illustrates the situation, where A and B are represented by a bit sequence, and stored in some place of a storage. If we read the bit sequence from left to right, then we see the image A. On the other hand, if we read it from right to left, then we see the image B. The procedure for constructing such a bit sequence from A and B is described as follows.

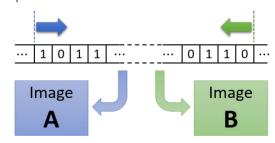


Figure 2: Illustration of an ambigramic bit sequence

Let us consider a typical case that aij and bij are 24-bit RGB color vectors as $a_{ij} = [r_{ij}^A, g_{ij}^A, b_{ij}^A]$ and b_{ij} $=[r_{ii}^B, g_{ii}^B, b_{ii}^B]$, each element of which is represented by 8 bits. Here, we introduce a binary representation of r_{ii}^A as $(r_{ij1}^A r_{ij2}^A r_{ij3}^A r_{ij4}^A | r_{ij5}^A r_{ij6}^A r_{ij7}^A r_{ij8}^A)_2$, where the third subscript in each element ranging from 1 to 8 indicates the significance of each digit, i.e., r_{ij1}^{A} denotes the most significant bit (MSB), and r_{ij8}^A denotes the least significant bit (LSB) in this binary representation. Other elements in a_{ij} and b_{ij} are also represented in binary in the same manner as r_{ij}^A . Moreover, we abbreviate a series of four bits $r_{ij1}^A r_{ij2}^A r_{ij3}^A r_{ij4}^A$ to $r_{ij1:4}^A$ for the sake of simplicity.

Author α σ ρ: Department of Communication Design Science, Kyushu University, 4-9-1, Shiobaru, Minami-ku, Fukuoka, 815-8540 Japan. e-mails: cuiwenyi1996@outlook.com, {k-inoue, hara}@design.kyushu-u.ac.jp

Our basic principle is to preserve more significant bits of the original image data in the resultant ambigramic image file. Following this principle, we first extract four MSBs from both A and B as follows:

$$r_{1,1,1:4}^{A}, g_{1,1,1:4}^{A}, b_{1,1,1:4}^{A}; \ r_{1,2,1:4}^{A}, g_{1,2,1:4}^{A}, b_{1,2,1:4}^{A}; \ \cdots \\ \cdots; r_{m,n,1:4}^{A}, g_{m,n,1:4}^{A}, b_{m,n,1:4}^{A} \quad (1) \\ r_{1,1,1:4}^{B}, g_{1,1,1:4}^{B}, b_{1,1,1:4}^{B}; \ r_{1,2,1:4}^{B}, g_{1,2,1:4}^{B}, b_{1,2,1:4}^{B}; \ \cdots \\ \cdots; \ r_{m,n,1:4}^{B}, g_{m,n,1:4}^{B}, b_{m,n,1:4}^{B} \quad (2)$$

where the horizontally adjacent pixels are separated by semicolons. Next, we reverse the order of the bit sequence in (2) as follows:

$$b_{m,n,4:1}^{B}, g_{m,n,4:1}^{B}, r_{m,n,4:1}^{B}; \cdots \cdots; b_{1,2,4:1}^{B}, g_{1,2,4:1}^{B}, r_{1,2,4:1}^{B}; b_{1,1,4:1}^{B}, g_{1,1,4:1}^{B}, r_{1,1,4:1}^{B}$$
 (3)

where note that the third subscripts (1:4) is also reversed as (4:1), e.g., $r_{1,1,4:1}^B$ denotes $r_{1,1,4}^B$, $r_{1,1,3}^B$, $r_{1,1,2}^B$, $r_{1,1,1}^B$. Then we combine (1) and (3) as follows:

$$(r_{1,1,1:4}^A|b_{m,n,4:1}^B)_2, (g_{1,1,1:4}^A|g_{m,n,4:1}^B)_2, (b_{1,1,1:4}^A|r_{m,n,4:1}^B)_2; \\ \cdots;$$

$$(r_{m,n,1:4}^{A}|b_{1,1,4:1}^{B})_{2},(g_{m,n,1:4}^{A}|g_{1,1,4:1}^{B})_{2},(b_{m,n,1:4}^{A}|r_{1,1,4:1}^{B})_{2} \tag{4}$$

which can be viewed as an ambigramic bit sequence, because if we read the bit sequence in (4) from left to right then the four MSBs in every pixel show the image A, on the other hand, if we read it from right to left then the four LSBs in every pixel are inversely read to show the image B. However, the resultant images can be corrupted by the replacement of the original four LSBs with others. To alleviate such a quality deterioration in the images, we next propose an error diffusion algorithm for generating better bit sequence, which improves the visual quality of the images.

The proposed error diffusion algorithm is described as follows. For one image A, all pixels are processed in a standard raster scan order, and at the same time, for another image B, all pixels are processed in the inverse raster scan order. For the first pixel in image A, the original color vector $a_{1,1}$ changes to

$$\tilde{\boldsymbol{a}}_{1,1} = \begin{bmatrix} (r_{1,1,1:4}^{A}|b_{m,n,4:1}^{B})_{2} \\ (g_{1,1,1:4}^{A}|g_{m,n,4:1}^{B})_{2} \\ (b_{1,1,1:4}^{A}|r_{m,n,4:1}^{B})_{2} \end{bmatrix}.$$
 (5)

Subtracting $\tilde{a}_{1,1}$ from $a_{1,1}$, we define an error vector by

$$e_{1,1}^A = a_{1,1} - \tilde{a}_{1,1},\tag{6}$$

which is diffused to unprocessed neighboring pixels as

$$a_{1+k,1+l} \leftarrow a_{1+k,1+l} + w_{k,l} e_{1,1}^A,$$
 (7)

Where k and l denote relative indices to access the neighboring pixels, and $w_{k,l}$ denotes error diffusion coefficients or error filter [5]. On the other hand, for the last pixel in image B, the original color vector $b_{\rm m,n}$ changes to

$$\tilde{\boldsymbol{b}}_{m,n} = \begin{bmatrix} (r_{m,n,1:4}^{B} | b_{1,1,4:1}^{A})_{2} \\ (g_{m,n,1:4}^{B} | g_{1,1,4:1}^{A})_{2} \\ (b_{m,n,1:4}^{B} | r_{1,1,4:1}^{A})_{2} \end{bmatrix}. \tag{8}$$

Algorithm 1 Constructing ambigramic bit sequence

```
procedure AMBIGRAMIC(A, B)
2:
                       A \leftarrow \text{zeros\_like}(A) \text{ where } A = [a_{ij}]
3:
                       \tilde{B} \leftarrow \text{zeros\_like}(B) \text{ where } B = [\boldsymbol{b}_{ij}]
                       m, n, \bot \leftarrow A.shape
4:
                       for i \leftarrow 1 to m do
5:
                                 i_R \leftarrow m + 1 - i
                                 for j \leftarrow 1 to n do
7:
8:
                                           j_R \leftarrow n + 1 - j
                                           Round the elements of a_{ij} and b_{i_R,j_R} to 8-bit
           integers
                                         \tilde{\boldsymbol{a}}_{ij} \leftarrow \begin{bmatrix} (r_{i,j,1:4}^{A} | b_{i_{R},j_{R},4:1}^{B})_{2} \\ (g_{i,j,1:4}^{A} | g_{i_{R},j_{R},4:1}^{B})_{2} \\ (b_{i,j,1:4}^{A} | r_{i_{R},j_{R},4:1}^{B})_{2} \end{bmatrix}
\tilde{\boldsymbol{b}}_{i_{R},j_{R}} \leftarrow \begin{bmatrix} (r_{i_{R},j_{R},1:4}^{B} | b_{i_{I},j,4:1}^{A})_{2} \\ (g_{i_{R},j_{R},1:4}^{B} | g_{i,j,4:1}^{A})_{2} \\ (b_{i_{R},j_{R},1:4}^{B} | r_{i,j,4:1}^{A})_{2} \end{bmatrix}
10:
11:
                                           e_{ij}^A \leftarrow a_{ij} - \tilde{a}_{ij}
12:
                                           oldsymbol{e}_{i_R,j_R}^B \leftarrow oldsymbol{b}_{i_R,j_R} - 	ilde{oldsymbol{b}}_{i_R,j_R} \ 	ext{for} \ (k,l) \in KL_{ij} \ 	ext{do}
13:
14:
                                                      \begin{aligned} & \boldsymbol{a}_{i+k,j+l} \leftarrow \boldsymbol{a}_{i+k,j+l} + w_{k,l} \boldsymbol{e}_{i,j}^{A} \\ & \boldsymbol{b}_{i_R-k,j_R-l} \leftarrow \boldsymbol{b}_{i_R-k,j_R-l} + w_{k,l} \boldsymbol{e}_{i_R,j_R}^{B} \end{aligned}
15:
16:
                                           end for
17:
                                 end for
18:
                       end for
19.
          return \tilde{A} = [\tilde{a}_{ij}] and/or \tilde{B} = [\tilde{b}_{i,j}]
           end procedure
```

Subtracting $\tilde{b}_{\text{m,n}}$ from $b_{\text{m,n}}$, we define an error vector by

$$\boldsymbol{e}_{m,n}^{B} = \boldsymbol{b}_{m,n} - \tilde{\boldsymbol{b}}_{m,n}, \tag{9}$$

which is diffused to unprocessed neighboring pixels as

$$\boldsymbol{b}_{m-k,n-l} \leftarrow \boldsymbol{b}_{m-k,n-l} + w_{k,l} \boldsymbol{e}_{m,n}^{B}, \tag{10}$$

where note that the sign of the relative indices k and l is reversed except in $w_{\rm k,l}$ because of the inverse raster scan.

After the above error diffusion procedures in (7) and (10), we proceed to the next pixels $a_{\rm 1,2}$ and $b_{\rm m,n-1}$, and the error diffusion procedures are repeated until the end of the scan.

The proposed error diffusion procedure is summarized in Algorithm 1, where the function 'zeros_like' returns an array of zeros being the same

size as the argument array, A.shape is a property to get the array dimensions of A, and KLii denotes a set of the pair of the relative indices k and l satisfying $1 \le i + k \le m$ and $1 \le i + l \le n$.

Experimental Results III.

we this section, demonstrate performance of the proposed method for formatting image data into an ambigramic data structure.

First, we show an example of the proposed ambigramic image data with two images selected from the standard image database SIDBA [6]. Figures 3(a) and (b) show the two standard images from SIDBA. Assume that Figs. 3(a) and (b) are images A and B, respectively. Then Algorithm 1 returns corresponding ambigramic images $\tilde{A} = [\tilde{a}_{ij}]$ and $\tilde{B} = [\tilde{b}_{ij}]$ as shown in Figs. 3(c) and (d), which can be saved as an ambigramic bit sequence, the forward reading of which shows Fig. 3(c), and the backford reading shows Fig. 3(d). For example, the RGB values at the top left pixel of Fig. 3(c) and at the bottom right pixel of Fig. 3(d) are shown in Fig. 4, where the top left and the bottom right pixels \tilde{a}_{11} and \tilde{b}_{mn} have the RGB values (232; 130; 116) and (46; 65; 23), which are also expressed as binary numbers. If we reverse the order of the bit sequences of RGB values in \tilde{a}_{ij} , then we have the bit sequences of BGR values in \tilde{b}_{mn} , and vice versa.



(a) Image A

(b) Image B





(c) \tilde{A} by Alg. 1

(d) \tilde{B} by Alg. 1

Figure 3: Original images (a) and (b) converted into ambigramic data (c) and (d)

R: 232 (1110 1000)₂ G: 130 (1000 0010), \widetilde{a}_{11} B: 116 (0111 0100),

R: 46 (0010 1110)₂ G: 65 (0100 0001)₂ B: 23 (0001 0111),

Figure 4: Comparison of pixel values

Figures 5(a) and (b) show two original images each of which has 640 × 480 pixels, and the following figures show the format-converted images from the original ones (a) and (b). Figures 5(c) and (d) show the graphics interchange format (GIF) images, and the GIF format is widely used on the Web due to its wide support and portability [4]. Figures 5 (e) and (f) show the results of the proposed ambigramic method without error diffusion (ED), where a daruma (dharma) doll and a beckoning cat can be seen as well as the above GIF images. Figures 5(g) and (h) show the results with ED (Algorithm 1), where we used Floyd and Steinberg's error filter [7] for $w_{k,l}$ in (7) and (10) as shown in Fig. 6 where "#" denotes the current pixel being processed, and "-" denotes the past pixel, and the visual quality is improved compared with the former results (e) and (f).





(a) Image A

(b) Image B





(c) GIF image A (d) GIF image B



(e) \tilde{A} without ED (f) \tilde{B} without ED



(g) \tilde{A} by Alg. 1 (h) \tilde{B} by Alg. 1

Figure 5: Examples of file format conversion

The GIF images are palette-based ones each of which has a palette table of 256 colors. That is, the GIF images are a sort of halftone images [5] as well as the proposed ambigramic images. To evaluate the quality of those images objectively and quantitatively, we present a parameter-free measure based on integral image [8].

-	#	$W_{0,1} = \frac{7}{16}$
$W_{1,-1} = \frac{3}{16}$	$W_{1,0} = \frac{5}{16}$	$W_{1,1} = \frac{1}{16}$

Figure 6: Error diffusion coefficients by Floyd-Steinberg

First, we compute the integral images of both reference (original) R and test (format-converted) T images, which are denoted as $\bar{R}=[\bar{r}_{ij}]$ and $\bar{T}=[\bar{t}_{ij}]$, respectively. Then we compute the signal-to-noise ratio (SNR) [9] of the integral images \bar{R} and \bar{T} as follows:

$$ISNR(R,T) = SNR(\bar{R},\bar{T})$$
(11)

$$= 10 \log_{10} \frac{\sum_{i=1}^{m} \sum_{j=1}^{n} \bar{r}_{ij}^{2}}{\sum_{i=1}^{m} \sum_{j=1}^{n} (\bar{r}_{ij} - \bar{t}_{ij})^{2}}.$$
 (12)

We call this measure the integral image-based SNR (ISNR).

For the format-converted images in Fig. 5, their ISNRs are plotted in Fig. 7, where the vertical and horizontal axes denote the ISNR value and image name, respectively, and green, yellow and orange bars denote the file formats, GIF, ambigramic format without ED and the final ambigramic format by Algorithm 1, respectively. Algorithm 1 achieves higher ISNR values than GIF

format for both images, which demonstrates the effectiveness of the proposed algorithm.

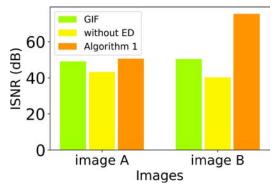


Figure 7: Integral image-based SNR (ISNR)

Additionally, we examine the applicability of the proposed algorithm to natural images shown in Fig. 8, where Fig. 8(a) shows five pairs of natural images where the top row shows the first images A, and the second row shows the corresponding second ones B. Figures 8(b), (c) and (d) show the corresponding format-converted images by GIF, the proposed ambigramic method without ED and Algorithm 1. Although, in Fig. 8(c), we can see false contours, they are removed in Fig. 8(d).

The ISNR values for those images in Fig. 8 are summarized in Table 1, where the proposed Algorithm 1 achieves higher values than the GIF images and ambigramic images without ED.

Table 1: ISNR values (dB)

	Pair 1	Pair 2	Pair 3	Pair 4	Pair 5
GIF image A	49.97	49.39	46.80	48.41	50.13
GIF image B	46.51	50.96	46.26	48.82	47.08
A without ED	38.05	46.97	33.59	47.34	45.28
B~ without ED	34.87	45.56	35.46	50.90	40.00
A~ by Alg. 1	73.93	60.81	82.48	95.21	75.76
B [~] by Alg. 1	63.61	88.92	53.09	94.37	85.13

IV. Conclusions

In this paper, we proposed an algorithm for formatting an ambigramic image file into which two images of the same size are recorded. If we read it forward direction, then we see the first image, on the other hand, if we read it backward direction, then we see the second one. We compared the proposed ambigramic image file format with the GIF format which has acquired a widespread use on the Web, and demonstrated that the proposed ambigramic images achieved higher quality than the GIF images based on an image quality measure. Experimental results showed that the proposed algorithm is also applicable to natural images.

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(a) Original images



(b) GIF images



(c) Ambigramic images without ED



(d) Ambigramic images by Algorithm 1

Figure 8: Examples with natural images

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Θ (1) Time Parallel Algorithm for Finding 2D Convex Hull on a Reconfigurable Mesh Computer Architecture

By Jelloul Elmesbahi, Mohammed Khaldoun, Omar Bouattane & Ahmed Errami

Hassan II University

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GJCST-F Classification: G.1.0



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Θ (1)Time Parallel Algorithm for Finding 2D Convex Hull on a Reconfigurable Mesh Computer Architecture

Jelloul Elmesbahi^a, Mohammed Khaldoun^a, Omar Bouattane^b & Ahmed Errami^a

Abstract- In this paper we propose a parallel algorithm in image processing in $\Theta(1)$ time, intended for a parallel machine "Reconfigurable Mesh Computer (RMC), of size n x n Elementary Processors (PE). The algorithm consists in determining the convex envelope of a two-level 2D image with a complexity in $\Theta(1)$ time. The approach used is purely geometric. It is based solely on the projection of the coordinates of PEs retained in specific quadrants and on the application of the algorithm that determines the Min / Max in $\Theta(1)$ time.

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I. Introduction

he problem of the convex envelope (convex hull) of a plane shape has been the subject of several studies in recent years. This very important problem concerns several areas such as image processing, pattern recognition or robotics [FU97]. The convex envelope can be used, for example, to normalize a shape, to triangulate a set of points, to extract topological characteristics, to decompose a complex shape in order to facilitate its recognition.

Compared to the very important number of works that have been realized on the use of sequential algorithms to extract the convex hull, the number of parallel algorithms to solve this problem is significantly lower. We have to wait until the beginning of the 80's to see the first parallel algorithms, which calculate the convex hull of a set of points [CHO81], [NAT81], [ALD83]. Since then, several authors have presented parallel algorithms to solve this problem with different levels of complexity [KIM87], [LIN93], [PRA89] and on different input data as for example on a set of points in a plane or in a space, on bi-level images, on multi-level images

Author α σ ω: NEST Research Group, LRI Laboratory E.N.S.E.M, Hassan II University, Casablanca, Morocco.

e-mails: j.elmesbahi@ensem.ac, aerrami@yahoo.fr,

m.khaldoun@ensem.ac.ma

Author p: ENSET Mohammedia, Hassan II University, Morocco etc.

e-mail: o.bouattane@gmail.com

with a single component or several components. On the other hand, the algorithms proposed in these works, were designed for different models of machines such as the model Mesh Connected Computer [MIL85], the model Polymorphic Torus [LI89] or the model Reconfigurable Mesh Computer [HAY98].

The synthesis of these different works leads us to note that the diversity of the parameters put into play complicates any attempt at global comparison between the performances of the different approaches. Table 1 presents a synthesis of some bibliographical references on this problem by specifying each time the machine model used, the nature of the input data as well as the degree of complexity obtained.

Among the parallel algorithms for determination of the convex hull of the connected components of a multi-level image, mention may be made of [ERR05], which describes an approach based on the structural characterization algorithm. As for the algorithm [BOU02] applies a purely geometric method. Our algorithm calculates with a complexity of Θ (1) the convex hull of an image or set of points with two levels of gray. This algorithm uses a purely geometric approach. It is based solely on the projection of the coordinates of PEs retained in specific quadrants and on the application of the algorithm that determines the Min / Max in Θ (1) time [ELMES91]. This allowed us to reduce the complexity of our solution compared to those proposed in the literature. The contribution of this document is summarized as follows:

- The proposed method guarantees constant time processing because all steps are executed in Θ (1) time, while most existing algorithms are based on image processing multiple times.
- The proposed provides the convex hull not only for a set of points but also for images with two gray levels.

Machine model Nature of data Complexity Bibliography Mesh connected Computer Points Sets θ(log n) [MIL85] O(1) Polymorphic Torus Points Sets [LI89] par ensemble de points θ(log n) [MIL88] hypercube Points Sets O(log n) [STO88] **Pyramid** Points Sets O(log n) [STO88] Pyramid tree Points Sets $\theta(\log^3 n/(\log\log n)^2)$ [MIL88] Mesh of tree Reconfigurable Mesh Points Sets $\theta(\log n)$ [MIL88] **EREW PRAM** Points Sets θ(log n) [MIL88] Reconfigurable Mesh Sorted Points Sets O(log log n)2 [HAY98] Bi-level image O(log²n) by component [MIL93] Reconfigurable Mesh one component O (log n)n is the number of Reconfigurable Mesh segments that approximate the Multilevel image the Computer contour of largest [BOU02] several components (RMC) component of the image. O (log n)n represents the total Reconfigurable Mesh number of extreme points Multilevel image Computer present on the contour of the [ERR05] several components (RMC) largest component of the image

Table 1: Synthesis of some works on the problem of the extraction of the convex hull

In this article, section 2 is devoted to the calculation model used for the implementation of the proposed algorithm. Section 3 presents the algorithm for determining the convex hull. Section 4 presents the result of determination of the convex hull of a two-level image matrix. This paper concludes with some remarks on future work.

II. Computational Model Architecture

The computational model of machine used in this paper is a parallel Reconfigurable Mesh Computer (RMC) [ELMES86] of size nxn Elementary Processors (PEs) arranged in a square matrix. It is a machine which respects the SIMD (Single instruction multiple data) model, in which each PE is located in the matrix by its row and column coordinates and can be characterized by its identifier ID = n.i + j, where i and j represent the row number and the column number respectively. In this architecture, Figure. 1 (a) refers to a set of PEs, each one is connected to its four neighbors; if they exist; through communication channels. It has a finite number of registers; of bus length log 2 (n) bits, in which it stores data to perform arithmetic and logic operations. All PEs can perform reconfiguration operations to exchange information with other PEs in the mesh. Figure 1. (b) describes the set of possible configurations of each PEThe communication of each PE with its neighbors is implemented through four different operations:

Single Bridge (SB): A PE of the RMC is in the SB state, if it establishes links between two of its communication channels, either in transmitting mode or the receiving one. In addition, it may disconnect some of its communication channels. Figure 1.b) shows the six possible configurations (1-6) of the SB configurations

Double Bridge (DB): A PE is in DB state, when it achieves configuration involving two independents buses. The configurations from 7 to 10 of Figure 1.b) shows the three possible cases of DB

Cross Bridge (CB): A PE goes to the CB state if it connects all its active communication channels into one. The figure 1 (b) shows the only possible configuration (11) with the CB operation.

Direct Broadcast: The direct broadcast operation consists in transmitting information from a transmitting PE to a set of receiving ones. The implementation of this operation is achieved as follows:

- a. All PEs are in CB mode. The receiving PEs are coupled by their receiving ports. While the transmitting PEs are in the transmitting mode via their ports.
- b. The transmitting PE sends the information throughout its ports. Then, all other PE receive the same information.

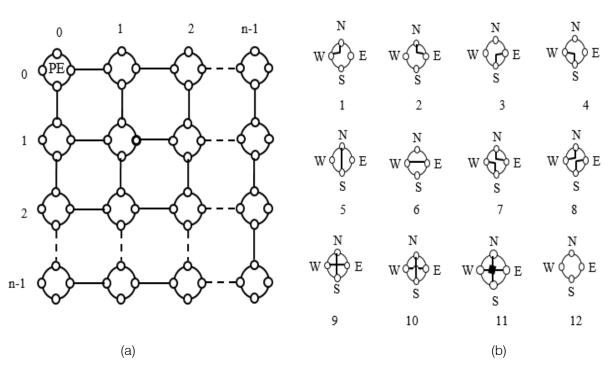
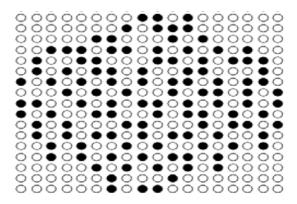


Figure 1: (a) Concerned model, (b) possible configurations

III. θ (1) Time Algorithm to Determine the Convex Hull of a Bi-level Image

Either to determine the convex hull of a set of points or PEs (Elementary Processors) arranged in the form of an image matrix as shown in Figure 2.



• Marked PEOUnmarked PE

Figure 2: Elementary Processors (PEs) arranged in the form of a set of points representing points of an image matrix.

The algorithm for determining the convex hull consists of 7 steps:

- 1. Step 1: Determination of points: i_m , i_M , j_m , j_M and points A, B, C, D, E, F, G and H (Figure 3).
- 2. Step2: Determination of points: $I_1^{'}$, $I_2^{'}$, $I_3^{'}$ et $I_4^{'}$: (Figure 10).
- 3. Step 3: Determination of points: I_1 , I_2 , I_3 et I_4 : (Figure 3).

- 4. Step 4: The delimitation of the area P_1 (quadrant: NW), then the determination of the PEs belonging to the convex hull forming part of this zone. The procedure will be applied to zones P_2 , P_3 and P_4 . (Figure 3).
- 5. Step 5: Labeling of all PEs belonging to the zones P₁, P₂, P₃and P₄ (Figure 8).
- 6. Step 6: Elimination of all PEs that do not belong to the zones P₁, P₂ P₃ and P₄. (Figure 8).
- 7. Step 7: Processing on the zones P1, P2, P3 and P4: this consists in determining PEs of the different quadrants NW, NE, SE and SW belonging to the convex envelope.

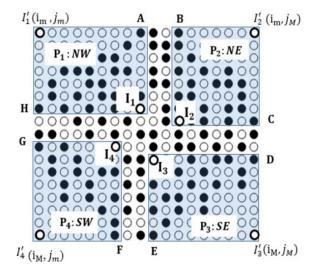


Figure 3: Step 1, Step 2 and Step 3.

Step 1:

- a) All the PEs of the image matrix get into Crosse Bridge (CB) (Figure 3) and perform in Θ (1) time [ELMES91]:
- 1. MIN operation for lines i. (i_m is the minimum of the
- MIN operation for columns j. (j_mbeing the minimum of the columns).
- MAX operation for lines i. (i_M is the maximum of the
- 4. MAX operation for columns j. (j_M is the minimum of the columns).



Figure 4 a: One PE in Crosse Bridge

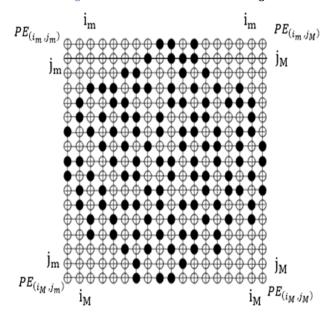
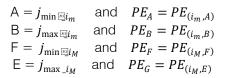


Figure 4 b: All the PEs of the image matrix in Crosse Bridge

Figure 4: Crosse Bridge

The (i_m, i_M, j_m, j_M) are all determined in $\Theta(1)$ time corresponding **PEs** $PE_{(i_m,j_m)},$ $PE_{(i_{m},j_{M})}, PE_{(i_{M},j_{m})}, PE_{(i_{M},j_{M})}$

b) All the PEs in each line go into Simple Bridge (SB) and make a MIN for the columns j then a Max for the columns j. We will have in each line at most two Marked PEs (PE_M) . (Figure 5-a). For example, for the line i_{m:}



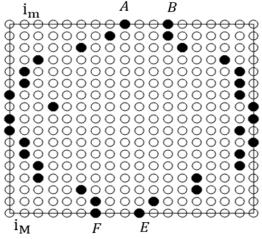


Figure 5 a: Min and Max lines

$$\begin{split} \mathbf{A} &= j_{\min \equiv i_m} \text{ ,B } = j_{\max \equiv i_m} \\ \mathbf{F} &= j_{\min \equiv i_M} \text{,E } = j_{\max _i_M} \end{split}$$

a) All the PEs of the image matrix get into Crosse Bridge (CB) (Figure 3) and perform in Θ (1) time [ELMES91]:

$$\begin{aligned} \mathbf{H} &= i_{\min_j_{\mathrm{m}}} & \text{ and } PE_{H} = PE_{(H,j_{\mathrm{m}})} \\ \mathbf{G} &= i_{\max_j_{\mathrm{m}}} & \text{ and } PE_{G} = PE_{(G,j_{\mathrm{m}})} \\ \mathbf{C} &= i_{\min_j_{\mathrm{M}}} & \text{ and } PE_{C} = PE_{(C,j_{\mathrm{M}})} \\ \mathbf{D} &= i_{\max_j_{\mathrm{M}}} & \text{ and } PE_{D} = D_{(D,j_{\mathrm{M}})} \end{aligned}$$

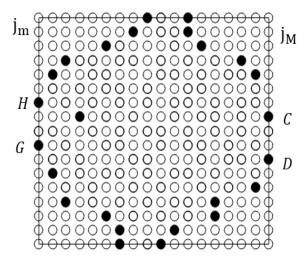


Figure 5 b: Min and Max columns

$$\begin{split} \mathbf{H} &= i_{\min_j_{\,\mathrm{m}}}, \ \mathbf{G} = \ i_{\max_j_{\,\mathrm{m}}} \\ \mathbf{C} &= i_{\min_j_{\,\mathrm{M}}}, \ \mathbf{D} = i_{\max_j_{\,\mathrm{M}}} \end{split}$$

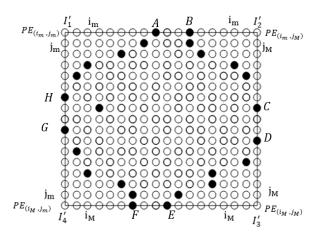


Figure 5: Summits: $PE_{(i_m,j_m)}$, $PE_{(i_m,j_m)}$, $PE_{(i_m,j_m)}$, $PE_{(i_m,j_m)}$ et $PE_M: M \in \{A,B,C,D,E,F,G,H\}$. belonging to the sides of the quadrilateral encompassing the set of points of the image.

Step 2:

- a) All PEs are configured in (SB) with respect to the line. (Figure 5).
- b) The $PE_{(i_m,j_m)}$, $PE_{(i_M,j_m)}$, transmit on the bus on their right a code indicating their presence. All unmarked PEs will receive this code, those to the right of the two PEs having (i_m,j_m) and (i_M,j_m) .
- c) The $PE_{(i_m,j_M)}$, $PE_{(i_M,j_M)}$ transmit on the bus to their left a code indicating their presence. All unmarked PEs will receive this code, those to the left of the two PEs having (i_m,j_M) and (i_M,j_M) .
- d) The operations indicated in (a), (b) and (c) will be done on the columns.
- e) Result: 4 PEs will receive 2 codes one on their line and another one on the columns and will memorize them in I'₁, I'₂, I'₃ and I'₄. (Figure 5).

Step 3:

a) All the PEs of the image matrix put themselves in (SB) on their column (Figure 6₋a).

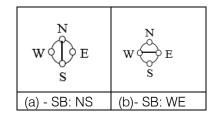


Figure 6: SB Rows and columns configuration (a) - SB: NS; (b)- SB: WE

b) The PE_A sends a code indicating the position of the column where it is located. So, all the PEs in this column will receive this code (Figure 7: quadrant P1). c) All PEs in the image matrix are configured as (SB) on their line (Figure6-b).

The PE_H sends a code to all the PEs that are on his line. So, all PEs on this line will receive this code (Figure 7: quadrant P1). This operation runs for all others PEs:PE_B,PE_C,PE_D,PE_F,PE_F and PE_G.

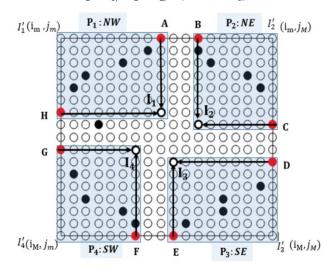


Figure 7: Determination of points I_1 , I_2 , I_3 and I_4 .

Step 4: Delimitation of the zones $P_1P_2P_3$ and P_4 : in the area P_1 (quadrant NW), all PEs that are between A and I_1 are in configuration (SB) on their column.

The PE_A gives them a code indicating that they have to block the bus that connects them with the PEs that are on their right (Figure 8.a).

In the same way PE_{I_1} communicates to the PEs that are in its line on the left, the code for disconnect PEs from the top line (Figure 8. a).

This operation will be executed in the zones $P_{2},\,P_{3}\,\mbox{and}\,\,\,P_{4.}$

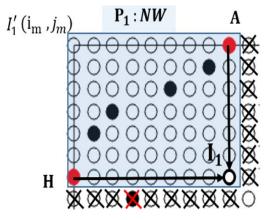


Figure 8 a: Delimitation of P₁ area.

Step 5: Labeling of the zones P_1 , P_2 , P_3 and P_4 : in zone P_1 , since the PEs I'_1 , A and I_1 are known, we proceed to the labeling all PEs belonging to P_1 .

All the PEs in zone P1 are configured in Crosse Bridge (CB) state. The PE_{I_1} transmits an identifier code of the zone P1(Figure 8.b).

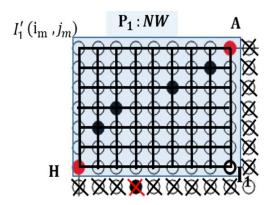


Figure 8b: Labeling of area P1

This operation will be executed in the zones P_2 , P_3 and P_4 (Figure 8).

Consequence: All PEs in these areas are identified as part of the zones P_1 , P_2 , P_3 and P_4

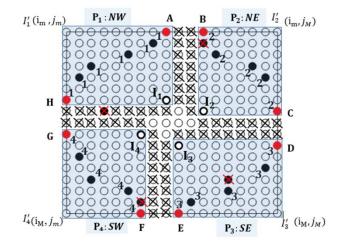


Figure 8: Delimitation and labeling of the zones P_1 , P_2 , P_3 and P_4

Step 6: Elimination of all PEs that do not belong to the zones P_1 , P_2 , P_3 and P_4 . (Figure 8).

- a) All the PEs in the image matrix go into state (CB) except the PE $\rm I'_{1.}$
- b) The PEI_1' transmits to all PEs of the image matrix the cancellation order of marking if it is a marked PE. This implies that all PEs that do not belong to zones P_1 , P_2 , P_3 and P_4 will no longer be marked. Those who belong to these areas remain marked if they are.
- Step 7: Treatment of labeled PEs belonging to zone P_i (Figure 9): Treatment of the area P₁ (Quadrant NW)

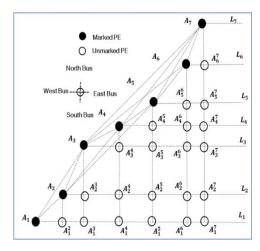


Figure 9: Treatment of the area P₁ (Quadrant NW)

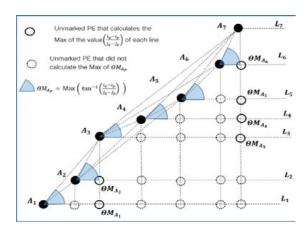


Figure 10: Determination of $Max \left(\frac{i_q - i_p}{j_q - j_p}\right)$

- a) In each column only one PE marked, all others must be eliminated. Any Marked PE that has a labelled PE with an index lower than its index in the same column must be discarded. (Example, see Figure 8, zone P_3).
- b) All unmarked PEs in the same column will execute the SB state except for only one remaining PE (A_7 , A_6 , A_5A_4 , A_3 , A_2 and A_1).
- c) Each marked PE transmits its position (i) on its south down column.
- d) Each marked PE transmits on the East bus its position (j).
- e) All the unmarked PEs A_p^q of row p and column q, which have received on the North and West buses the positions i_p and j_q , calculate the value $\binom{i_q-i_p}{j_q-j_p}$

in
$$(A_1^2, ..., A_1^7, A_2^3, ..., A_2^7, A_3^4, ..., A_5^7, ..., A_5^7, et A_6^7)$$
.

- $\left(\frac{i_q-i_p}{j_q-j_p}\right)$ or EachPEA_P calculate: $\tan^{-1}\left(\frac{i_q-i_p}{j_q-j_p}\right)$, all the unmarked PEs of the line p execute the operation Max [ELMES91] in this line.
- 1. In the case of the line L_1 all $PE_1(A_1^2,...,A_1^7)$ will run the Max's $\left(\frac{i_q-i_p}{j_q-j_n}\right)$ or $\theta M_{A_p}=$ Max (θ_p^q) and we are going to have A_1^3 who will have the Max of $\left(\frac{i_q-i_p}{i_q-i_p}\right)$.
- In the line L_2 the PE A_2^3 will have the Max.
- In the line L_3 the PE A_3^7 will have the Max.
- In the line L_4 the PE A_4^7 will have the Max.
- 5. In the line L_5 the PE A_5^7 will have the Max.
- In the line L_6 the PE A_6^7 will have the Max.

All unmarked PEs, having already calculated execute a Max [ELMES91] of the value $i = i_q - i_p$ on their column j_q .

Example:

- 1. In the line L_1 , A_1^3 will be selected (it has the value $(i_3 - i_1)$ which is greater than $(i_3 - i_2)$
- 2. In the line, A_3^7 will be selected (it has the value $(i_7 - i_3)$ which is greater than $(i_7 - i_4) >$ $(i_7 - i_5) > (i_7 - i_6)$.
- 3. L₁, A_1^3 transmits through the open bus to the PE A_1^2 the order to eliminate the $PE_M(A_2)$ on its column.
- L_3 , A_3^7 transmits through the open bus to all PEs $A_3^4 A_3^5$ and A_3^6 the order to eliminate all PE_M on their column. In the other lines, $noPE_M$ will be selected.

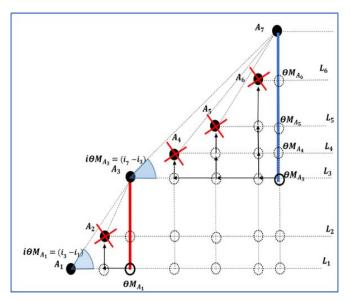


Figure 11: Max on the line index on the same column

Result: The Marked PEs ((PE) M): A 2, A 4, A (5) and A (6) are eliminated, the [PE] MA1, A3 and A (7)belong to the convex hull of the area

Conclusion IV.

This paper has dealt with a parallel algorithm of determining the convex hull of a two-level 2D image with a complexity Θ (1) time. This is executed on a parallel machine (RMC), of size n x n Elementary Processors. The algorithm is essentially based on projections and the calculation of Min / Max in Θ (1) time. In a future work, we will extend this algorithm to 3D space and always in Θ (1) time.

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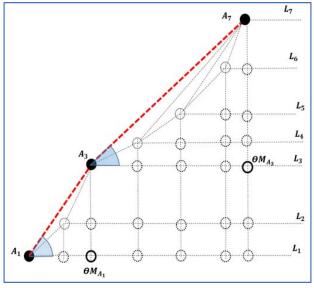


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Texture Classification of 3D Mr Color Images using 3D Orthogonal Rank Filters

By Arun Kumar A & E. G. Rajan

University of Mysore

Abstract- The term 'texture' refers to patterns arranged in an order in a line or a curve. Textures allow one to make a meaningful interpretation of certain geometric regularity of spatially repeated patterns. In addition, texture also exhibits useful information about spatial distribution of color or gray intensities in an image. Correct interpretation of latent textures of various tissues in a body is an important requirement for a surgeon as a preoperative measure. In this context, extraction of textures in an MR scanned 3D image would assist a medical professional in the preoperative decision making process. This paper proposes a novel technique for extracting directional textures of a 3D MR image in all three axes separately.

Keywords: 3D color images, superficial and volumetric features, texture classification.

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Texture Classification of 3D Mr Color Images using 3D Orthogonal Rank Filters

Arun Kumar A a & E. G. Raian a

Abstract- The term 'texture' refers to patterns arranged in an order in a line or a curve. Textures allow one to make a meaningful interpretation of certain geometric regularity of spatially repeated patterns. In addition, texture also exhibits useful information about spatial distribution of color or gray intensities in an image. Correct interpretation of latent textures of various tissues in a body is an important requirement for a surgeon as a preoperative measure. In this context, extraction of textures in an MR scanned 3D image would assist a medical professional in the preoperative decision making process. This paper proposes a novel technique for extracting directional textures of a 3D MR image in all three axes separately.

Keywords: 3D color images, superficial and volumetric features, texture classification.

I. Introduction

his paper describes a computationally efficient technique to detect various texture characteristics as directional features in a given 3D digital image. The computational tool used for this purpose is '3D Rank Filters', which are essentially directional filters. These filters cause radical changes in the original content of a given image but precisely extract various textures.

Any given 3D MR image consists of texture features of tissues corresponding to muscle fibers in almost all directions. One can visualize major muscle fibers of a body component with naked eye. But most of the finer textures cannot be visualized even by an expert, in which case machine vision support system becomes guite handy. The algorithms presented in this paper could be used to detect texture patterns in all the three orthogonal axes of a 3D rectangular discrete coordinate system in which 3D digital image is displayed.

II. LITEERATURE SURVEY

Apart from detecting latent textures in a given image, one can also artificially create texture images. Fig. 1 shows a 3D texture image, which is artificially generated using a cellular automaton rule.

Author a: Department of Computer Science University of Mysore, Manasa Gangotri Mysore, Karnataka, India.

e-mail: arun.arigala@gmail.com

Author o: Director, Rajiv Gandhi International School of Information Technology, MG-MIRSA Approved Research Centre of Mysore University. e-mail: rajaneg@yahoo.co.in

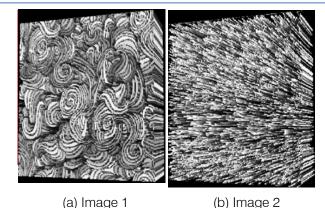
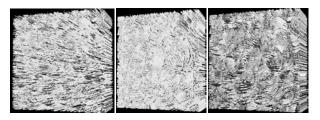


Figure 1: Texture images due to cellular automaton rules

Fig. 2 shows texture patterns extracted from image 2 shown in Fig. 1 along three axes of 3D rectangular coordinate system.



Textures along x axis Textures along y axis Textures along z axis Figure 2: Textures along all axes of 3D coordinate system

Two texture features are usually considered for image segmentation. They are (i) spatial frequency features and (ii) average gray level features. Either 'structural approach' or 'statistical approach' could be used for developing texture detection algorithms. Mostly statistical approach is considered for texture classification because of ease in parametrization and quantification of texture features.

Edge detection is a method by which one would be able to detect edge pixels details which help determine characteristics of texture complexities. For instance, directions of edges could be treated as characteristics of textures in determining patterns in the textures.

Consider a region with N pixels in a given image. Any gradient-based edge detector algorithm could be applied to this region, which would yield two outputs for every pixel p, viz, 'gradient magnitude Mag(p)' and 'gradient direction' Dir(p). Now, the edgeness per unit area of a given image is defined by $|\{p|Mag(p) > T\}|$

the expression

for some predefined

threshold T. Let $H_{maq}(R)$ be the normalized histogram of the gradient magnitudes of the region of interest R, and let H_{dir}(R) be the normalized histogram of the gradient orientations of the region of interest R. Both are normalized according to the size $N_{\textrm{R}}$ Then, one can define a quantitative measure $F_{maq,dir} = (H_{maq}(R), H_{dir}(R))$ for describing texture of the region of interest R.

Another technique to quantify texture is 'cooccurrence matrix', which defines features of a texture using certain spatial relations of similar gray values. Such numerical features could be used for texture classification. Some of the standard features from a normalized co-occurrence matrix are given below.

Angular moment =
$$2^{\text{nd}}$$
 $\sum_{i} \sum_{j} p[i,j]^2$

Contrast =
$$\sum_{i=1}^{Ng} \sum_{j=1}^{Ng} n^2 p[i,j], ext{ where } |i-j|=n$$

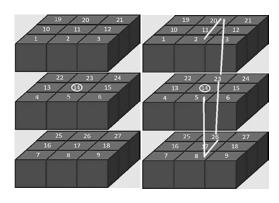
Correlation =
$$rac{\sum_{i=1}^{Ng}\sum_{j=1}^{Ng}(ij)p[i,j] - \mu_x\mu_y}{\sigma_x\sigma_y}$$

$${\sf Entropy} = \qquad \qquad -\sum_i \sum_j p[i,j] ln(p[i,j])$$

where p[i, j]_is the [i, j]th entry in a gray-level spatial dependence matrix, and Ng is the number of grayvalues in the quantized image. It is to be noted that the co-occurrence matrix based feature extraction will not yield comfortable visual perception.

III. Proposed Method

As outlined earlier, the term 'textures' refers to 'repeated patterns' in a given image. Consider the 27neighborhood window shown in Fig. 3. The cells 1, 2, 3, 4, 5, 6, 7, 8, 9 form the first plane, 10, 11, 12, 13, 14, 15, 16, 17, 18 the middle plane and cells 19, 20, 21, 22, 23, 24, 25, 26, 27 form the rear plane of the window. The given 3-D digital image is plane-wise raster-scanned by this window (See Fig. 3). In order to extract 3-D linear textures along an axis with a directional twist, one has to choose that particular axis and its associated rank of a particular directional twist. For example if one chooses the X axis and rank1 of zero directional twist, values in cells 2,11,20,23,26,17,8,5 would be read and stored in an array. The reading pattern is shown in Fig. 4.



- (a) Labeled window
- (b) Reading direction

Figure 3: 27-neighborhood window in layer form and the reading direction

The values of the cells 2,11,20,23,26,17,8,5 are the boundary values corresponding to the central voxel 14. The plane formed by these cells is perpendicular to the X axis as shown in Fig. 4.

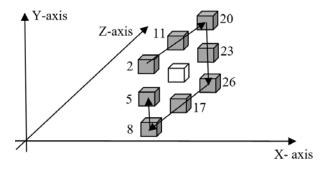


Figure 4: 3D coordinate axes and the reading plane

X-axis rank 2 consists of cells 11, 20, 23, 26, 17, 8, 5, 2 and the corresponding plane is perpendicular to X axis as given in Fig. 4 but with a directional twist of 45 degrees. One can construct four ranks in X-axis, four in Y-axis and another four in Z-axis as shown in Table 1. A total of 12 rank filters could be constructed in three axes which are called "3D Orthogonal Rank Filters".

Table 1: Ranks filters in all three axes

Axes	Ranks	Cell sequences		
X	X1	2,11,20,23,26,17,8,5		
	X2	11,20,23,26,17,8,5,2		
^	Х3	20,23,26,17,8,5, 2,11		
	X4	23,26,17,8,5, 2,11,20		
	Y1	4,13,22,23,24,15,6,5		
Y	Y2	13,22,23,24,15,6,5,4		
Y	Y3	22,23,24,15,6,5,4,13		
	Y4	23,24,15,6,5,4,13,22		
	Z1	10, 11,12,15,18,17,16,13		
Z	Z2	11,12,15,18,17,16,13,10		
	Z3	12,15,18,17,16,13,10, 11		
	Z4	15,18,17,16,13,10, 11,12		

IV. Texture Classification of 3d Medical **I**MAGES

Textures of a medical image play an important role in support of a surgeon to decide the angle at which the surgical blade should be used to make incision so that the loss of blood due to surgery is kept minimum. A case study was carried out to verify the validity of the algorithm and the result of the study presented in Fig. 5, which is self-explanatory.

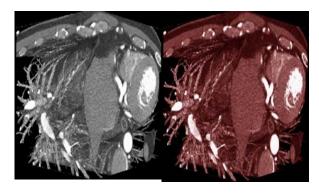


Figure 5: Sample MRI image and its colored version

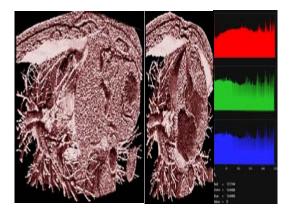


Figure 6: Rank x1 filtered, sectioned (30-140) and histogram

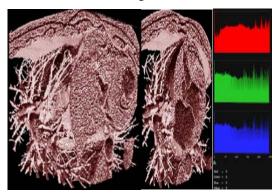


Figure 7: Rank x2 filtered, sectioned (30-140) and histogram

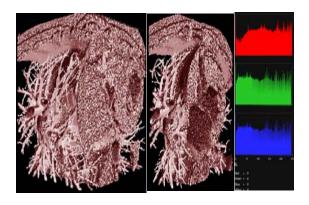


Figure 8: Rank x3 filtered, sectioned (30-140) and histogram

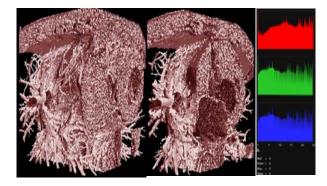


Figure 9: Rank x4 filtered, sectioned (30-140) and histogram

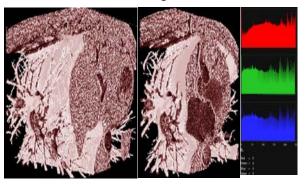


Figure 10: Rank y1 filtered, sectioned (30-140) and histogram

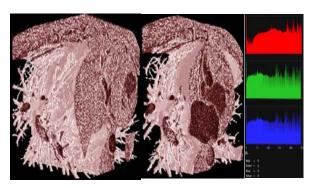


Figure 11: Rank y2 filtered, sectioned (30-140) and histogram

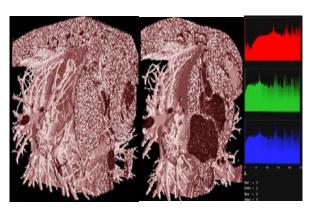


Figure 12: Rank y3 filtered, sectioned (30-140) and histogram

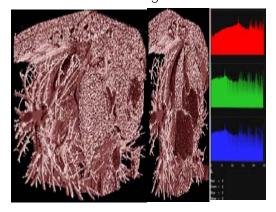


Figure 13: Rank y4 filtered, sectioned (30-140) and histogram

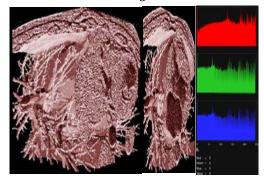


Figure 14: Rank z1 filtered, sectioned (30-140) and histogram

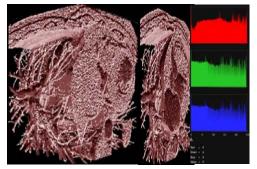


Figure 15: Rank z2 filtered, sectioned (30-140) and histogram

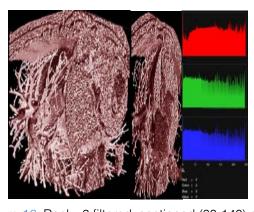


Figure 16: Rank z3 filtered, sectioned (30-140) and histogram

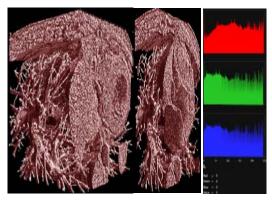


Figure 17: Rank z4 filtered, sectioned (30-140) and histogram

V. STATISTICAL RESULTS

Original 3D image statistics

Pixels Count 568089
Pixels without black 449360
Red Min 0
Red Max 252

Red Mean 90.0212677943069 Red Standard Deviation 65.4300367403954

Red Median 92
Red Total Count 568089
Green Min 0
Green Max 249

Green Mean 48.2296154299766 Green Standard Deviation 56.6653938791041

Green Median 33
Green Total Count 568089
Blue Min 0
Blue Max 249

Blue Mean 48.2296154299766 Blue Standard Deviation 56.6653938791041

Blue Median 33
Blue Total Count 568089
Saturation Min 0
Saturation Max 1

Saturation Mean 0.347210377454758

Saturation Standard Devia	ation 0.225063592195511	Cb Max WB	-0.00196078419685364
Saturation Median	0.341176480054855	Cb Mean WB	-0.0371856279671192
Luminance Min	0	Cb Standard Deviation WE	
Luminance Max	0.980392158031464	Cb Median WB	-0.0411764681339264
Luminance Mean	0.270397782325745	Cr Min WB	0.00196081399917603
	ation 0.234511524438858	Cr Max WB	0.123529434204102
Luminance Median	0.243137255311012	Cr Mean WB	0.10223163664341
Y Min	0.243137233311012	Cr Standard Deviation WB	
Y Max	0.976470589637756		
Y Mean	0.236716106534004	Cr Median WB	0.115686297416687
Y Standard Deviation	0.227980986237526	Rank X1 filtered image sta	
Y Median	0.196078434586525	Pixels Count	563200
Cb Min	-0.0450980365276337	Pixels without black	441456
Cb Max	-0.00196078419685364	Red Min	0
Cb Mean	-0.0298237334936857	Red Max	252
Cb Standard Deviation	0.0168844126164913	Red Mean	122.136278409091
Cb Median	-0.0372548997402191	Red Standard Deviation	87.516547830135
Cr Min	-0.00196078419685364	Red Median	121
Cr Max	0.123529434204102	Red Total Count	563200
Cr Mean	0.0804557129740715	Green Min	0
Cr Standard Deviation	0.0502428002655506	Green Max	249
Cr Median	0.103921592235565	Green Mean	88.18859375
Red Min WB	1	Green Standard Deviation	
Red Max WB	252	Green Median	57
Red Mean WB	113.806507032224	Green Total Count	563200
Red Standard Deviation W		Blue Min	0
Red Median WB	111	Blue Max	249
Red Total Count WB	449360	Blue Mean	88.18859375
Green Min WB	0	Blue Standard Deviation	85.5399836766261
Green Max WB	249	Blue Median	57
Green Mean WB	60.972747908136	Blue Total Count	563200
Green Standard Deviation	WB 57.2920439382876	Saturation Min	0
Green Median WB	48	Saturation Max	1
Green Total Count WB	449360	Saturation Mean	0.30152890086174
Blue Min WB	0	Saturation Standard Devia	
Blue Max WB	249	Saturation Median	0.317647069692612
Blue Mean WB	60.972747908136	Luminance Min	0
Blue Standard Deviation V		Luminance Max	0.980392158031464
Blue Median WB	48	Luminance Mean	0.411772221326828
Blue Total Count WB	449360		ation 0.335566163063049
Saturation Min WB	0.235294118523598	Luminance Median	0.34901961684227
Saturation Max WB	1	Y Min	0
Saturation Mean WB	0.43894961476326	Y Max	0.976470589637756
	tion WB0.154169782996178	Y Mean	0.384158581495285
Saturation Median WB	0.396078437566757	Y Standard Deviation	0.334127157926559
Luminance Min WB	0	Y Median	0.298039227724075
Luminance Max WB	0.980392158031464	Cb Min	-0.0450980365276337
Luminance Mean WB	0.341841757297516	Cb Max	-0.00196078419685364
	ation WB0.212376952171326	Cb Mean	-0.0245320294052362
Luminance Median WB	0.309803932905197	Cb Standard Deviation	0.0166890006512403
Y Min WB Y Max WB	0 0.976470589637756	Cb Median	-0.0254901945590973
Y Mean WB	0.299260765314102	Cr Min	-0.00196078419685364
Y Standard Deviation WB	0.216774046421051	Cr Max	0.123529434204102
Y Median WB	0.258823543787003	Cr Mean	0.0646790862083435
Cb Min WB	-0.0450980365276337	Cr Standard Deviation	0.0495649240911007
ON WILL TAD	5.0 100000021 0001	Cr Median	0.0686274766921997

Red Min WB	1	Green Standard Deviation	90 97/19/600109/
Red Max WB	252	Green Median	94
Red Mean WB	155.818817730419	Green Total Count	559680
Red Standard Deviation W		Blue Min	0
Red Median WB	150	Blue Max	249
Red Total Count WB	441456	Blue Mean	109.47983133219
Green Min WB	0	Blue Standard Deviation	89.8741846221284
Green Max WB	249	Blue Median	94
Green Mean WB	112.509097169367	Blue Total Count	559680
Green Standard Deviation			
Green Median WB	88	Saturation Min	0
Green Total Count WB	441456	Saturation Max	1
Blue Min WB	0	Saturation Mean	0.275802910327911
Blue Max WB	249	Saturation Standard Devia	
Blue Mean WB	112.509097169367	Saturation Median	0.30588236451149
Blue Standard Deviation W		Luminance Min	0
Blue Median WB	88	Luminance Max	0.980392158031464
Blue Total Count WB	441456	Luminance Mean	0.489122450351715
			ation 0.351963937282562
Saturation Min WB	0.235294118523598	Luminance Median	0.486274510622025
Saturation Max WB	1	Y Min	0
Saturation Mean WB	0.384684056043625	Y Max	0.976470589637756
	tion WB0.142558693885803	Y Mean	0.463929653167725
Saturation Median WB	0.329411774873734	Y Standard Deviation	0.35080423951149
Luminance Min WB	0	Y Median	0.439215689897537
Luminance Max WB	0.980392158031464	Cb Min	-0.0450980365276337
Luminance Mean WB	0.525330066680908	Cb Max	-0.00196078419685364
Luminance Standard Devia	ation WB0.289833068847656	Cb Mean	-0.0224275775253773
Luminance Median WB	0.4666666668653488	Cb Standard Deviation	0.015973724424839
Y Min WB	0	Cb Median	-0.02156862616539
Y Max WB	0.976470589637756	Cr Min	-0.00196078419685364
Y Mean WB	0.490101218223572	Cr Max	0.123529434204102
Y Standard Deviation WB	0.300843000411987	Cr Mean	0.0583211965858936
Y Median WB	0.415686279535294	Cr Standard Deviation	0.0472652688622475
Cb Min WB	-0.0450980365276337	Cr Median	0.0607843399047852
Cb Max WB	-0.00196078419685364	Red Min WB	1
Cb Mean WB	-0.0307566896080971	Red Max WB	252
Cb Standard Deviation WE	3 0.013269835151732	Red Mean WB	178.322465743467
Cb Median WB	-0.0333333313465118	Red Standard Deviation W	/B 61.7170538248096
Cr Min WB	0.00196081399917603	Red Median WB	204
Cr Max WB	0.123529434204102	Red Total Count WB	440208
Cr Mean WB	0.0830569192767143	Green Min WB	0
Cr Standard Deviation WB	0.039645180106163	Green Max WB	249
Cr Median WB	0.0960784554481506	Green Mean WB	139.192545342202
Rank Y1 filtered image state	tistics	Green Standard Deviation	
Pixels Count	559680	Green Median WB	163
Pixels without black	440208	Green Total Count WB	440208
Red Min	0	Blue Min WB	0
Red Max	252	Blue Max WB	249
Red Mean	140.256889651229	Blue Mean WB	139.192545342202
Red Standard Deviation	91.295496046247	Blue Standard Deviation V	
Red Median	155	Blue Median WB	163
Red Total Count	559680	Blue Total Count WB	440208
Green Min	0	Saturation Min WB	0.235294118523598
Green Max	249	Saturation Max WB	1
O M	100 47000100010	Caturation Maan MD	0.050655505000775

Saturation Mean WB

0.350655525922775

Green Mean

109.47983133219

Saturation Standard Devia	tion WB0.105345770716667	Y Mean	0.391311198472977
Saturation Median WB	0.321568638086319	Y Standard Deviation	0.319555670022964
Luminance Min WB	0	Y Median	0.3333333343267441
Luminance Max WB	0.980392158031464	Cb Min	-0.0450980365276337
Luminance Mean WB	0.621869742870331	Cb Max	-0.00196078419685364
LuminanceStandardDevia	tion WB0.273765563964844	Cb Mean	-0.0252390317618847
Luminance Median WB	0.717647075653076	Cb Standard Deviation	0.0168824307620525
Y Min WB	0	Cb Median	-0.0294117629528046
Y Max WB	0.976470589637756		
Y Mean WB	0.589839696884155	Cr Min	-0.00196078419685364
		Cr Max	0.123529434204102
Y Standard Deviation WB	0.286698818206787	Cr Mean	0.066443957388401
Y Median WB	0.686274528503418	Cr Standard Deviation	0.0499764792621136
Cb Min WB	-0.0450980365276337	Cr Median	0.0803921818733215
Cb Max WB	-0.00196078419685364	Red Min WB	1
Cb Mean WB	-0.0279822442680597	Red Max WB	252
Cb Standard Deviation Wi		Red Mean WB	162.220466007254
Cb Median WB	-0.0254901945590973	Red Standard Deviation W	B59.5814731225976
Cr Min WB	0.00196081399917603	Red Median WB	163
Cr Max WB	0.123529434204102	Red Total Count WB	440594
Cr Max WB	0.123529434204102	Green Min WB	0
Cr Mean WB	0.0746816620230675	Green Max WB	249
Cr Standard Deviation WB	0.0398297160863876	Green Mean WB	116.723936322328
Cr Median WB	0.0686274766921997	Green Standard Deviation	
Rank Z1 filtered image sta	tistics	Green Median WB	104
Pixels Count	573516	Green Total Count WB	440594
Pixels without black	440594	Blue Min WB	0
Red Min	0	Blue Max WB	249
Red Max	252	Blue Mean WB	116.723936322328
Red Mean	124 623138674422	Blue Standard Deviation W	B 73 1980388120368
Red Mean Red Standard Deviation	124.623138674422 86.0969667214582	Blue Standard Deviation W	
Red Standard Deviation	86.0969667214582	Blue Median WB	104
Red Standard Deviation Red Median	86.0969667214582 130	Blue Median WB Blue Total Count WB	104 440594
Red Standard Deviation Red Median Red Total Count	86.0969667214582 130 573516	Blue Median WB Blue Total Count WB Saturation Min WB	104 440594 0.235294118523598
Red Standard Deviation Red Median Red Total Count Green Min	86.0969667214582 130 573516 0	Blue Median WB Blue Total Count WB Saturation Min WB Saturation Max WB	104 440594 0.235294118523598 1
Red Standard Deviation Red Median Red Total Count Green Min Green Max	86.0969667214582 130 573516 0 249	Blue Median WB Blue Total Count WB Saturation Min WB Saturation Max WB Saturation Mean WB	104 440594 0.235294118523598 1 0.354308485984802
Red Standard Deviation Red Median Red Total Count Green Min Green Max Green Mean	86.0969667214582 130 573516 0 249 89.6711966187517	Blue Median WB Blue Total Count WB Saturation Min WB Saturation Max WB Saturation Mean WB Saturation StandardDeviati	104 440594 0.235294118523598 1 0.354308485984802 on WB0.123913042247295
Red Standard Deviation Red Median Red Total Count Green Min Green Max Green Mean Green Standard Deviation	86.0969667214582 130 573516 0 249 89.6711966187517 80.8826837693388	Blue Median WB Blue Total Count WB Saturation Min WB Saturation Max WB Saturation Mean WB Saturation StandardDeviati Saturation Median WB	104 440594 0.235294118523598 1 0.354308485984802 on WB0.123913042247295 0.317647069692612
Red Standard Deviation Red Median Red Total Count Green Min Green Max Green Mean Green Standard Deviation Green Median	86.0969667214582 130 573516 0 249 89.6711966187517 80.8826837693388 66	Blue Median WB Blue Total Count WB Saturation Min WB Saturation Max WB Saturation Mean WB Saturation StandardDeviati Saturation Median WB Luminance Min WB	104 440594 0.235294118523598 1 0.354308485984802 on WB0.123913042247295 0.317647069692612 0
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Table 2: Comparison og selected statistical parameters

Selected Statistical Parameter	Original Image	Rank X1 Filtered Image	Rank Y1 Filtered Image	Rank Z1 Filtered Image
Red Min	0	0	0	0
Red Max	252	252	252	252
Red Mean	90.02	122.13	140.25	124.62
Red Standard Deviation	65.43	87.51	91.29	86.09
Red Median	92	121	155	130
Red Total Count	568089	563200	559680	573516
Green Min	0	0	0	0
Green Max	249	249	249	249
Green Mean	48.22	88.18	109.47	89.67
Green Standard Deviation	56.66	85.531	89.87	80.88
Green Median	33	57	94	66
Green Total Count	568089	563200	559680	573516
Blue Min	0	0	0	0
Blue Max	249	249	249	249
Blue Mean	48.22	88.18	109.47	89.67
Blue Standard Deviation	56.66	85.53	89.87	80.88
Blue Median	33	57	94	66
Blue Total Count	568089	563200	559680	573516

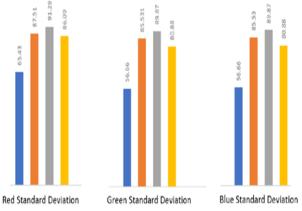


Figure 18: Graph showing textures statistics

VI. Conclusion

From the experimental study, it was observed that the variations in the statistical parameter values remain almost uniform, especially standard deviations of

Red, Green and Blue values. One can infer similar behavior as far as other parameters also.

All four texture versions of the image obtained using rank filters could be seen to provide a visual proof of the fact textures in an image are direction sensitive and so they could be used for image segmentation purposes.

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The Impact of Different Image Thresholding based Mammogram Image Segmentation- A Review

By Krishnaveni

Abstract- Images are examined and discretized numerical capacities. The goal of computerized image processing is to enhance the nature of pictorial data and to encourage programmed machine elucidation. A computerized imaging framework ought to have fundamental segments for picture procurement, exceptional equipment for encouraging picture applications, and a tremendous measure of memory for capacity and info/yield gadgets. Picture segmentation is the field broadly scrutinized particularly in numerous restorative applications and still offers different difficulties for the specialists. Segmentation is a critical errand to recognize districts suspicious of tumor in computerized mammograms. Every last picture have distinctive sorts of edges and diverse levels of limits. In picture transforming, the most regularly utilized strategy as a part of extricating articles from a picture is "thresholding". Thresholding is a prevalent device for picture segmentation for its straightforwardness, particularly in the fields where ongoing handling is required.

GJCST-F Classification: I.2.10



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The Impact of Different Image Thresholding based Mammogram Image Segmentation-A Review

Krishnaveni

Abstract- Images are examined and discretized numerical capacities. The goal of computerized image processing is to enhance the nature of pictorial data and to encourage programmed machine elucidation. A computerized imaging framework ought to have fundamental segments for picture procurement, exceptional equipment for encouraging picture applications, and a tremendous measure of memory for capacity and info/yield gadgets. Picture segmentation is the field broadly scrutinized particularly in numerous restorative applications and still offers different difficulties for the specialists. Segmentation is a critical errand to recognize districts suspicious of tumor in computerized mammograms. Every last picture have distinctive sorts of edges and diverse levels of limits. In picture transforming, the most regularly utilized strategy as a part of extricating articles from a picture is "thresholding". Thresholding is a prevalent device for picture segmentation for its straightforwardness, particularly in the fields where ongoing handling is required. Thresholding is the procedure of creating uniform locales taking into account the edge esteem. It is the methodology of utilizing a limit to concentrate the Region of Interest (ROI). In this paper we have talked about numerous edge choice routines, for example, histogram, clustering, entropy, object attribute, spatial and local methods are exceptionally valuable systems and the acquired results are analyzed in an indicated way. Accordingly Image transforming applications are exhibit in all spaces.

Introduction

igital Image Processing is a quickly advancing field with developing applications in science and Digital image processing is engineering [3]. adaptable research in this period [1]. Scientific visualization is the representation of data graphically as a means of gaining understanding and insight into the information. A wonderful place to start out learning scientific visualization is within the field of image process, since it involves algorithms that facilitate convert information into pictures. In today's technologyoriented world, the term 'image process' usually refers to the processing of a two-dimensional information set employing a computer [7]. Digital image processing involves the control and investigation of images or pictures utilizing digital computers [5]. Alongside the advancement of data innovation with the development of information technology (IT), computerized sign loaded with the entire world, so see the picture changed over to be computer to manage an advanced sign. Advanced picture transforming is through computer instrument, with computerized picture motion by a progression of handling operations, and get individuals with the needs of the application [4].

Many researchers implement differing types of organizations like image restoration. image improvement, color image process, image segmentation etc. Image improvement technique is among the only and most appealing space of digital image process. Improvement techniques like intensity conservation, distinction improvement highlight sure options means that rely that a part of the image wish to be enhance some application some input image as well as noise, reduction or removal of noise is additionally style of image improvement. Brightness preservation has increased visual quality of digital image in order that the limitation contained in these pictures is employed for varied applications during a higher method. A really common technique for image improvement is histogram equalization (HE) and curvelet transformation. HE technique is often utilized for image improvement owing to its simplicity and relatively higher performance on the majority forms of pictures. Another wide used technique is curvelet transformation. This system is known and separate bright regions of image however additional error rate and low Peak Signal to Noise Ratio (PSNR), result of this system is brightness preservation level is low and output image is grey [1].

Digital image process has several applications in several fields like medication, forensic, robotics, industrial automatic scrutiny systems, navigation etc. This field has attracted attentions of researchers and students to develop and/or to enhance algorithms for applications [2]. With of image process techniques, individuals will simply tamper digital pictures by usina some

Author: e-mail: krishnaveni120691@gmail.com

software system. For pictures are wide used for the recent years, great amount of digital image manipulation might be seen in magazine, Industry, Scientific Journals, Court Rooms, News etc. The tampered pictures can turn out nice impact, and hurt to the traditional order of the society. The way to build effectively forensics to the tampered pictures is changing into a hunt hotspot within the data security field. Wherever digital image forensics has emerged as a replacement analysis field that aims to reveal meddling in digital pictures detection forgery in digital pictures is a rising analysis field [6].

a) Thresholding

Thresholding could be a common image segmentation methodology that converts a gray-level image into a binary image. The choice of optimum thresholds has remained a challenge over decades [9]. Binarization (i.e., image thresholding) is wide used as a preprocess algorithmic rule in image analysis and understanding [17]. Image thresholding (or binarization) could be a basic kind of image segmentation capability [19]. In all ancient segmentation schemes, statically measured thresholds or primary points are wont to binarize pictures. Due to the variations in pictures characteristics, these techniques could generate high segmentation accuracy for a few pictures and low accuracy for different pictures. For many pictures, the quantity of grey level is way smaller than the quantity of pixels [15]. Intelligent segmentation by "dynamic" determination of thresholds supported image properties is also a lot of sturdy answer [18].

Thresholding is a crucial method in several image process applications [10] [13]. However, the execution time needs should still be important, particularly if it's of interest to perform period of time thresholding of an outsized variety of pictures, like within the case of high-resolution video sequences [10]. The image thresholding drawback is treated as a crucial issue in image process, and it can not only reduce the image data, however additionally lay a decent foundation for succedent target recognition and image sympathetic. Nature of global thresholdina segmentation and local thresholding was analyzed in image segmentation [8] [14]. In image analysis, image thresholding that is employed for separating the object from the background is one in every of the foremost common application. For the preprocessing functions of a picture, thresholding could be a necessary tool [16] [18].

Automatic thresholding is a very important technique within the image segmentation method. The essential plan of automatic thresholding is to mechanically choose an optimal gray-level threshold value for partitioning pixels within the pictures into object and background supported their gray-level distribution [12]. Entropy-based image thresholding has received

wide interest in recent years. It's a very important concept within the space image segmentation. The entropy-based approach was wont to get the brink of image from eighty ages; it's wont to weight the quantity of reserved data of image once segmentation [11]. Thresholding segmentation may be a vital preprocessing tread on several image process applications. However, most of the prevailing thresholding ways will solely cope with a picture with some special histogram patterns [13].

Examples of thresholding applications are document image analysis wherever the goal is to extract written characters [26], [27] logos, graphical content, musical scores, map process wherever lines, legends, characters are to be found [28], scene process wherever a target is to detected [29], quality examination of materials [30], [31]. Alternative applications embrace cell pictures [32], [33] and data illustration [34], segmentation of assorted image modalities for non-destructive testing applications, like ultrasonic pictures in [35], eddy current pictures [36], thermal pictures [37], X-raying computed tomography (CAT) [38], optical device scanning confocal research [38], extraction of edge field [39], image segmentation normally [40], [41] spatio-temporal segmentation of video pictures [42] etc.

II. LITERATURE REVIEW

Despite a lots of works out there within the literature, a handful of important explore works are reviewed here. In recent years, the outstanding advances in medical imaging instruments have enlarged their use significantly for medical specialty likewise as designing and follow-up of treatment [20]. Thresholding is that the simplest technique of image segmentation. From a grayscale image, thresholding is wont to produce binary pictures (Shapiro, et al. 2001:83) [21].

Martin Luessi et al.. discussed image thresholding could be a quite common image process operation, since the majority image process schemes would like some style of separation of the pixels into totally different categories. So as to work out the thresholds, most ways analyze the histogram of the image. The optimum thresholds are usually found by either minimizing or maximizing an objective function with reference to the values of the thresholds. By process two categories of objective functions that the optimum thresholds may be found by efficient algorithms [22].

Bong Chin-Wei et al analyses thresholding may be a common region segmentation technique. During this technique a threshold is chosen, and a picture is split into collections of pixels having value but the threshold and collections of pixels with values bigger or adequate to the brink. In 2007, Nakid and his team

have planned to use the multi-objective approach to find the optimal thresholds of three criteria: the within-class criterion, the entropy and therefore the overall chance of error criterion [23].

There are varieties of survey papers on thresholding. Lee, Chung, and Park [44] conducted a comparative analysis of five global thresholding techniques and advanced helpful criteria thresholding performance analysis. In an earlier work, Weszka and Rosenfeldcon jointly outlined many analysis criteria [45]. Palumbo, Swaminathan and Srihari addressed the problem of document binarization compares three techniques; whereas Trier and Jain had for most in depth comparison basis techniques in the context of character segmentation from complicated backgrounds [46]. Sahoo et al analyses nine thresholding algorithms and illustrated relatively their performance [47]. Glasbey have introduced the relationships and performance variations between eleven histogram-based algorithms supported an in depth statistical study [48].

Kapur et al (1985) employed the Global entropic thresholding algorithm. Unsupervised thresholding progress wherever the most excellent thresholding grev level is chosen by exhaustive search among obtainable grey intensities has been improved. One of the approach examined by the author is the make use of signal dispensation methods specifically thresholding and information fusion to recover the correctness of information mined from the restructured tomograms (Mwambela & Johansen 2001, Mwambela 1999, Mwambelaet al 1997) [24]. Murthy et al have demonstrated the use of fuzzy and rough set theories to grip the vagueness there in pictures whereas performing histogram thresholding. Pal et al in the year 1983 established make use of the concept of decreasing fuzziness measures, which enumerate vaqueness in information to achieve image segmentation based on histogram thresholding [25].

Solihin and Leedham have developed a global thresholding technique to extract written components from low-quality documents [59]. In an additional motivating approach Aviad and Lozinskii [60] have pioneered semantic thresholding to emulate human approach to image binarization. The "semantic" threshold is found by minimizing measures of conflict criteria in order that the binary image resembles most to a "verbal" description of the scene. Gallo and Spinello [61] have developed a method for thresholding and isocontour extraction via fuzzy arithmetic. Fernandez [62]has investigated the choice of a threshold in matched filtering applications within the detection of tiny target objects. During this application the Kolmogorov-Smirnov distance between the background and object histograms is maximized as a purpose of the threshold value.

Anderson, J. et al have propose a technique supported the graph cut thresholding method, that is all the same acceptable for hardware (FPGA) time period implementations. The image of the weld pool was processed employing a series of methods: image truncation, bi-level thresholding, median filter and edge detection. Recently, a bi-level image thresholding technique supported graph cut was projected. The technique provided thresholding results that were superior to those obtained with previous techniques. Moreover, the technique was computationally less complicated compared to different graph cut-based image thresholding approaches. However, the execution time necessities should still be vital, particularly if it's of interest to perform time period thresholding of about sized range of pictures, like within the case of highresolution video sequences, [80].

Traditional best thresholding techniques are terribly computationally high once extended to multilevel thresholding for their thoroughly search mode. Thus their applications are restricted. One in every of the foremost well-liked techniques for image segmentation is understood as multilevel thresholding. Multilevel thresholding amounts to segmenting a gray-level image into many distinct regions. The most distinction between multilevel and binary thresholding, is that the binary thresholding outputs a two-color image, sometimes black and white, whereas the multilevel thresholding outputs a gray scale image within which a lot of details from the first image may be unbroken. Two major issues with utilizing the multilevel thresholding technique are: it's a time overwhelming approach, i.e., finding acceptable threshold values may take exceptionally long process time; process a correct range of thresholds or levels that may keep most of the relevant details from the first image may be a troublesome task [81].

Existing Image Thresholding III. **TECHNIQUES**

The output of the thresholding operation could be a binary image whose grey level of zero (black) can indicate a picturing element fit in to a print, legend, drawing, or target and a grey level of one (white) can indicate the background. Taxonomy of thresholding algorithms supported on the sort of knowledge used. We have a tendency to distinguish six classes, namely, thresholding algorithms supported the exploitation of 1) Histogram entropy data, 2) Histogram shape data, 3) Image attribute data like contours, 4) Clump of graylevel data, 5) Domestically adaptative characteristics, 6) Spatial data [43].

- 1. Histogram shape-based techniques wherever the peaks, valleys and curvatures of the ironed histogram are measured and analyzed.
- Clustering-based techniques wherever the grey level 2. samples are clustered in two components as

background and foreground (object) or alternately are measure shapely as two Gaussian distributions.

- Entropy-based techniques lead to algorithms, as an example, that uses the entropy foregroundbackground regions, the cross-entropy between the first and binarized image etc.
- Object attribute-based techniques measure of similarity between the gray-level and binarized pictures, like as fuzzy similarity, shape, edges, variety of objects etc.
- The spatial techniques use the likelihood mass performs models taking under consideration correlation between pixels on a global scale.
- Local techniques don't verify an only single value of threshold however adapt the threshold value relying upon the local image characteristics.

Histogram Shape-Based Thresholding Methods

This class of techniques achieves thresholding supported the form properties of the histogram. Essentially two most important peaks and an intervening valley is searched for using such tools because the protrusive hull of the histogram, or its curvature and 0 (zero) crossings of the wavelet elements. Alternative authors try and approximate the histogram via two-step functions or two-pole autoregressive smoothing.

Using a differencing operation on the ironed kernel, the histogram is characterized by the set S of peaks, that's the triplet of early, peaking and terminating zero-crossings on the peak detection $S = [(e_i, m_i, s_i), i = 1,...I]$, wherever I is that the variety of peaks wanted. The particular variety of peaks obtained is reduced to I, that's two for binarization, by adjusting the support of the smoothing filter and a peak-merging criterion. For two-level illustration of a picture the threshold ought to be somewhere in between the primary early and therefore the second terminating zero crossing, that is [50]:

$$T_{opt} = \gamma e_1 + (1 - \gamma) s_2, \quad 0 \le \gamma \le 1.$$

$H_f(T) = -\sum_{g=0}^T \frac{p(g)}{P(T)} \log \frac{p(g)}{P(T)} \text{ and } H_b(T) = -\sum_{g=T+1}^G \frac{p(g)}{P(T)} \log \frac{p(g)}{P(T)} \text{ one has [56]}:$

$$T_{ont} = \arg \max[H_f(T) + H_b(T)]$$

Yen, Chang and Chang [56] have thought about a multilevel thresholding method wherever additionally to the category entropies a cost purpose based on the amount of bits required to the thresholded image is enclosed.

Thresholding Algorithms Based on Attribute Similarity

The calculations considered under this class select the limit quality in light of some similitude measure between the first picture and the binarized

b) Clustering based thresholding methods

In this category of algorithms the grey level information undergoes a clump analysis with the amount of clusters being set to two. Alternately the grey level distribution is shapely as a combination of two Gaussian distributions representing, correspondingly, background and foreground regions.

Otsu advised minimizing the weighted total of within-class variances of the foreground associated background pixels to determine an optimum threshold. Since step-down of within-class variances is equal to the maximization of between-class scatter, the selection of the optimum threshold may be developed as [51]:

$$T_{opt} = \arg \max[P(T).(1 - P(T)).(m_f(T) - m_b(T))^2]$$

The Otsu technique provides satisfactory results once the numbers of pixels in every category are nearone other. The Otsu technique still remains one in every of the foremost documented thresholding techniques. During a similar study thresholding supported on isodata clump is given in Velasco [52]. Some limitations of the Otsu technique is mentioned in Lee [53].

c) Entropy based thresholding methods

This category of algorithms exploits the entropy of the distribution of the grey levels during a scene. The maximization of the entropy of the thresholded image is understood as indicative of most data transfer. Alternative authors try and minimize the cross-entropy between the input gray-level image and therefore the output binary image as indicative of preservation of data. Johannsen and Bille [54] and Pal, King, Hashim [55] were the primary to check Shannon entropy based mostly thresholding.

In this technique the foreground background categories are thought about as two completely different sources. Once the total of the two category entropies may be a most the image is alleged to be optimally thresholded. Therefore using the description of the foreground and background entropies,

adaptation of the picture. These characteristics can take the manifestation of edges, shapes, or one can specifically consider the first dim level picture to parallel picture similarity. Then again they consider certain picture properties, for example, reduction or integration of the items coming about because of the binarization process or the happenstance of the edge fields.

Hertz and Schafer [82] consider a multi thresholding method where a beginning global threshold assessment is refined provincially by considering edge data. The system expect that a diminished edge field is gotten from the dim level picture E_{grav}, which is contrasted and the edge field got from the binarized picture, E_{binary} (T). The edge is balanced in such a path, to the point that the fortuitous event between theories two edge fields is expanded. This infers there is least stipend for both overabundance edges and missed edges. For our situation we have considered a streamlined adaptation methodology. Both the dark level picture edge field and the twofold picture edge field have been gotten through the Sobel administrator. The worldwide limit is given by that esteem that expands the occurrence of the two edge fields in light of the check of coordinating edges and punishing the overabundance unique edges and the abundance thresholded picture edges.

$$T_{opt} = \arg \max[E_{gray} \cap E_{binary}(T)]$$

In a corresponding study Venkatesh and Rosin [83] have identified the difficulty of best possible thresholding for edge field assessment.

e) Spatial thresholding methods

In this category of algorithms one utilizes spatial details of object and background pixels, for instance, within the sort of context possibilities, correlation functions, co-occurrence possibilities, local linear dependence models of pixels, two-dimensional entropy etc. One in the entire primary to explore spatial details was Rosenfeld [63] who thought about such ideas as local average grey level for thresholding. Alternative authors have used relaxation to improve on the binary map [64], [65], the Laplacian of the images to enhance histograms [49], the quad tree thresholding and second-order statistics [66]. Co-occurrence probabilities have been used as indicator of spatial dependence as in Lie [67], Pal [68], and Chang [69]. Recently Leung and Lam have thought about thresholding within the context of a posteriori spatial chance estimation [70].

Chanda and Majumder [71] had advised the employment of co-occurrences for threshold choice. Lie [67] has projected many measures to the present result. Within the technique by Chang, Chen, Wang and Althouse the co-occurrence possibilities of each the initial image and of the thresholded image are calculated. A suggestion that the thresholded image is most kind of like the initial image is obtained whenever they possess as similar co-occurrences as doable. In alternative words the threshold T is set in such a way that the grey level transition possibilities of the initial image has minimum relative entropy (discrepancy) with reference to that of the initial image. This assess of similarity is obtained by the relative entropy, as an alternative referred to as the directed divergence or the Kullback-Leibler distance, that for two generic

distributions p, q has the shape
$$D(p,q) = \sum_{p,q} p \log \frac{p}{q}$$
.

Think about the four quadrants of the cooccurrence matrix: The primary quadrant denotes the background-to-background (bb) transitions whereas the third quadrant corresponds to the foreground-toforeground (ff) transitions. Equally the second and fourth quadrants denote, correspondingly, the background-toforeground (bf) and also the foreground-to-background (fb) transitions. Belongings the cell possibilities be denoted as p_{ii}, that is that the i to j grey level transitions normalized by the overall variety of transitions. The probabilities are

$$P_{bb}(T) = \sum_{i=0}^{T} \sum_{j=0}^{T} p_{ij} \; , \; P_{bf}(T) = \sum_{i=0}^{T} \sum_{j=T+1}^{G} p_{ij} \; , \; \; P_{ff}(T) = \sum_{i=T+1}^{G} \sum_{j=T+1}^{G} p_{ij} \; , \label{eq:Pbb}$$

$$P_{fb}(T) = \sum_{i=T+1}^{G} \sum_{j=0}^{T} p_{ij}$$
 and equally for the thresholded

image one finds the quantities $Q_{bb}(T)$, $Q_{bf}(T)$, $Q_{ff}(T)$, Q_{fd}(T). Plugging these expressions of co-occurrence possibilities within the relative entropy expression one will establish an optimum threshold as [69]:

$$\begin{split} T_{opt} = argmin[P_{bb}(T)logQ_{bb}(T) + P_{bf}(T)logQ_{bf}(T) + P_{ff}(T)logQ_{ff}(T) + P_{fb}(T)logQ_{fb}(T)] \\ T(i,j) = m(i,j) + [1 + k.(\frac{\sigma(i,j)}{R} - 1)] \end{split}$$

Locally adaptive thresholding strategies

A threshold that's calculated at every picture element characterizes this category of algorithms. The worth of the threshold depends upon some narrow statistics like vary, variance, and surface fitting parameters or their logical mixtures. It's typical of domestically adaptive strategies to own many adjustable parameters [72]. The threshold T (i, j) are going to be indicated as a purpose of the coordinates i. j; otherwise the thing or background selections at every picture element are going to be indicated by the logical variable B (i, j). Nakagawa and Rosenfeld [73], Deravi and Pal [74] were the first users of adaptive techniques for thresholding.

This technique claims to recover on the Niblack technique particularly for stained and badly well-lighted documents. It adapts the threshold according to the local mean and variance over a window size of bxb. The threshold at picture element (i,j) is calculated as:

where m(i,j) and $\sigma(i,j)$ are as in Niblack [59] and

Sauvola suggests the values of k=0.5 and R=128. Therefore the contribution of the standard deviation is converted into adaptive. For instance within the case of text written on a grimy or stained paper the threshold is down [75].

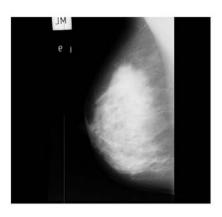
Among different local thresholding strategies specifically meshed to document pictures one will mention the work of Kamada and Fujimoto [76]who

develop a two-stage technique, the primary being a global threshold, followed by a neighborhood refinement. Eikvil, Taxt and Moen [77] think about a quick adaptive technique for binarization of documents whereas Pavlidis [78] uses the second-derivative of the gray-level image. Zhao and Ong [79] have thought about validity-guided fuzzy c-clustering to supply thresholding strong against illumination and shadow effects.

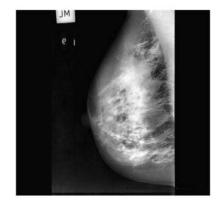
IV. RESULTS AND DISCUSSION

NORMAL Mammogram images of (Mdb003)

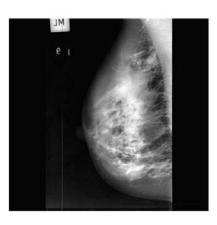
Histogram Shape based methods



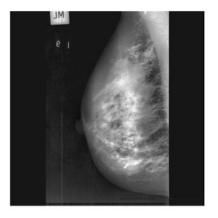
(a) Mdb003 — Original Image



(b)Mdb003 'Uniform'— Flat histogram



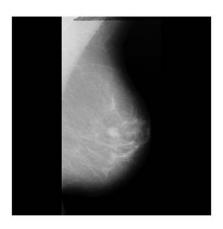
(c) Mdb003 'Rayleigh' — Bell-shaped histogram



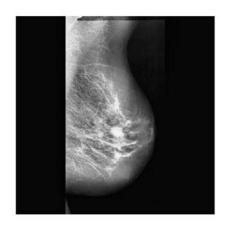
(d) Mdb003'Exponential' — Curved histogram

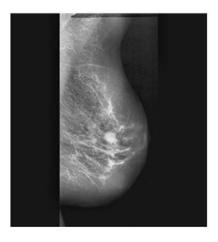
BENIGN Mammogram images of (Mdb010)

Histogram Shape based methods

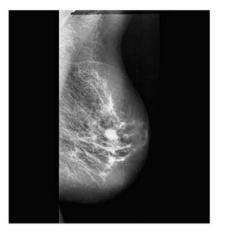


(a) Mdb010— Original Image





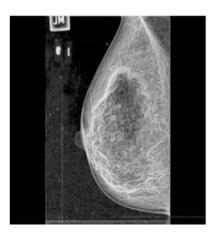
(c) Mdb010'Rayleigh' — Bell-shaped histogram

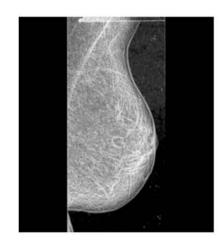


(d)Mdb010 'Exponential' — Curved histogram

Mammogram images of (Mdb003), (Mdb010) and (Mdb058)

Entropy based methods





(a) Mdb003 — Entropy based

(b) Mdb010 — Entropy based

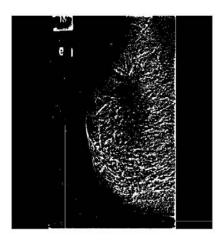


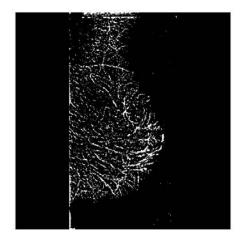
(c) Mdb058 — Entropy based

Mammogram images of (Mdb003), (Mdb010) and (Mdb058) **Spatial based methods** (a) Mdb003 — Spatial based (b)Mdb010 — Spatial based (c)Mdb058 — Spatial based

Mammogram images of (Mdb003), (Mdb010) and (Mdb058)

Locally Adaptive based methods





(a) Mdb003 — Locally Adaptive based

(b)Mdb010 — Locally Adaptive based



(c)Mdb058 — Locally Adaptive based

V. Examinations

Beside an unpleasant portrayal of every system, we introduce a valuable measurement and exchanges about the recurrence of the most utilized picture transforming techniques as a part of the issue of tiny

picture division. This investigation is useful for a superior utilization of existing systems, for enhancing their execution and in addition for outlining new ones. Table 1 demonstrates the most essential image thresholding systems found in the considered papers.

Author Name	Year	Domain	Model	Applications
Bamford and Lovell	1998	Cell Segmentation	Level set methods	Biological images
Solorzano et al.,	1999	Networking	World Wide Lightning Location Network (WWLN)	Lightning data for hurricanes
Cong and Parvin	2000	Segmentation and Classification	Image analysis techniques (the geometrical model fitting)	Cellular images
Boland and Murphy	2001	Pattern classification	Interpretation the concavity points	Microscope images
Malpica and de Solorzano	2002	Segmentation	Grey Level thresholding	Cellular images
Hu, et al.,	2004	segmentation	improved active contour model	Cell images.
Wahlby, et al.,	2004	segmentation	watershed segmentation	Cell images.
Naik, et al.,	2007	segmentation	Bayesian classifier and a level-set	Medical images
Lebrun et al.,	2007	segmentation	support vector machine (SVM)	Cellular images
Colantonio et al.,	2007	segmentation	fuzzy c-means algorithm	Medical image
Yang et al.,	2005	segmentation	gradient vector	Color images
Nilsson &Heyden,	2005	segmentation	level set methods and the watershed	Bone marrow sample images
Wang, et al.,	2008	Segmentation	Adaptive thrsholding algorithm	Leaf images
Angulo	2008	Segmentation	watershed segmentation and thresholding	Light channel image
Bai, et al.,	2009	Segmentation	Thresholding	MRI brain images
Coelho, et al.,	2009	Segmentation	watershed	Microscope Cell images
Dalle, et al.,	2009	Histopathology Image Segmentation	Thresholding	Histopathological H & E Stained Breast Cancer Images
Danek et al.,	2009	segmentation	graph-cut	Cellular images
Russell, et al.,	2009	segmentation	Stable Count Thresholding (SCT)	Cellular images
Ta, et al.,	2009	segmentation	Otsu's method	fluorescence
Zhou,et al.,	2009	segmentation	The adaptive thresholding and watershed, Markov model.	microscopic images Satellite imagery
Jeong, et al.,	2009	Classification	Thresholding	Microscopy images.
(Yang &Choe,	2009)	segmentation	graph-cut	Microscopy images.
Xiangzhi,et al.,	2009	Edge detection	Thresholding	Real time images
Madhloom,et al.,	2010	segmentation	The adaptive thresholding	Cellular images
Wei,et al.,	2011	segmentation	Renyi entropy thresholding	3-d images
Seroussi,et al.,	2012	Segmentation	Modified active contour model	Microscopy images
Ali El-Zaart and Ali A.Ghosn	2013	Segmentation	Bimodal and multimodal thresholding	MRI Brian images

Jin LIU	2014	Segmentation	3-d histogram based thresholding method	Two synthetic aperture radar (SAR) images and two license plate images
Temitope Mapayi et al.,	2015	Retinal Vessel Segmentation	Adaptive Thresholding Technique	Retinal image
James R. Parker	2015	Segmentation	Gray level thresholding	Various areas of the image
Akshay Upadhyay and Ramgopal Kashyap	2016	Segmentation	Intensity and Texture Based Segmentation	Medical Image
Murat Karakoyun et al	2017	Segmentation	Multilevel Thrsholding using Otsu's method	Real Images
K.P.Baby Resma et al	2018	Segmentation	Multilevel Thrsholding using Kapur and Otsu technique	Real Images
Hemeida. A. M.et al	2019	Segmentation	Multilevel Thrsholding	Standard Test Images

As pointed out in [Malpica and de Solorzano, 2002], the most widely spread segmentation method is grey level thresholding.

VI. CONCLUSIONS

Since there is no general methodology for getting precise picture segmentation, pretty much all systems consolidate the two fundamental methodologies: region based plans and edge based plans. This is way a characterization taking into account the paradigm utilized by every segmentation procedure is practically inconceivable. Rather, a rundown of the most utilized routines and how they are normally joined to accomplish great segmentation results is useful for better utilization of existing strategy and for enhancing their execution and in addition for planning new ones. In this paper we generally depict some illustrative studies in the field of thresholding for picture segmentation. Some of them utilize just basic transforming methods yet the larger part consolidates techniques without considering their multifaceted nature, e.g. edge with molecule calculation (Wang et al., 2008) or fuzzy cmeans calculation with manufactured neural system (Colantonio et al., 2007). As a general propensity we can presume that the new systems utilize two principle headings which appear to give steady and precise segmentation results. The first has a tendency to utilize the geometrical properties as from the earlier information, i.e. geometrical model fitting. At the point when this is unrealistic because of powerless limits, low between item complexities or high variability fit as a fiddle and size, the second inclination taking into account items gimmicks is viewed as; these peculiarities

are utilized to prepare an ANN, a Bayesian systems or a SVM.

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Computer Generation of Fractals Some Methods and Techniques

By Poli Venkata Subba Reddy

Sri Venkateswara University

Abstract- Fractals are objects, have the property of a system scale invariance or self-similarity. These objects are accursed in nature in the form of cost, hills, clouds, and waling act. These structures are need computer assistance for a generation. In this paper, the fractal dimensions were studied. Some methods and techniques are studied to simplify the Computer generation of fractals. Python programming is given to generate fractal graphics.

Keywords: fractals, fractal generation, fractal dimension, iteration, recursion, parallelism, python.

GJCST-F Classification: I.3.6



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Computer Generation of Fractals Some Methods and Techniques

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Abstract- Fractals are objects, have the property of a system scale invariance or self-similarity. These objects are accursed in nature in the form of cost, hills, clouds, and waling act. These structures are need computer assistance for a generation. In this paper, the fractal dimensions were studied. Some methods and techniques are studied to simplify the Computer generation of fractals. Python programming is given to generate fractal graphics.

Keywords: fractals, fractal generation, fractal dimension, iteration, recursion, parallelism, python.

I. Introduction

ractal structures of nature [1] are similar to themselves on different length-scales of observation. This geometrical property is studied for a great variety of irregular shapes, many of which result from the growth process.

Fractals geometry is introduced by Mandelbrot [I] as" the Geometry of Nature". Clouds are not spheres, mountains are not cones, coastlines are not circles and nor does lightning travel in a straight line, says Dr. Beroit Mandelbrot.

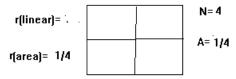
Human expertise is not sufficient to generate Fractals of nature. Fractals need computer assistance. Even Computer programming also requires methods and techniques to generate fractals. In the following, some methods and techniques are proposed for Computer generation of fractals.

II. Fractals and Dimension

Fractal is defined as similar to themselves of geometrical shapes. For instance, coastlines, mountains, rivers, etc. The fractal structures are studied through the fractal dimension and defined by

D = logN(h)/log(1/h)

Where h is the length of line-segment and N(h) is the number of line-segments.



D=log N/ log (1/r) =log 4/log2=2

Figure 1: Fractal scaling

Author: Department of Computer Science and Engineering, College of Engineering, Sri Venkateswara University, Tirupati, India. e-mail: pvsreddy@hotmail.co.in

III. Methods and Techniques

Computer generation of fractal shell be simplified by introducing methods and techniques. In the following, three methods and techniques are introduced to simplify the complexity of Computer generation of fractals.

a) Iteration

Iteration is the method in which output function value may be taken an input value to the function. This method is proposed to reduce the complexity of Computer generation of fractals.

This is given by

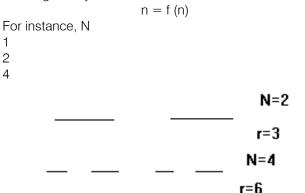


Figure 2: Iteration

$$D = \log(2/3) = 0.6309$$

Here, the number of self similarities can be defined as N = f(N).

b) Recursion

Recursion is a process that calls itself, directly or indirectly. This method can be applied to simplify the complexity of Computer generation of fractals using programming.

For instance, consider the generation of the Koch curve. The recursion method is applied to call self-similarity.



Figure 3: Recursion

$D = \log 4/\log 3 = \log 16/\log 9 = .26$

c) Parallel fractals

The parallel processing technique is divide number of sub-task of the task and each task will processed independently with individual processors in the Multiprocessing computer system. This parallel processing technique is proposed for Computer generation of fractals when the large number of computations and having the number of sub-tasks. The computer generation of fractals, in which the fractal can be divides into the number of sub-tasks, and each sub-task will be processed with independent processor and generate the self-similarities.

For instance, consider the Sierpinski gasket in which the triangle is divided into three triangles and each triangle will self-similarly generate with independent processor in Multiprocessing computer system.

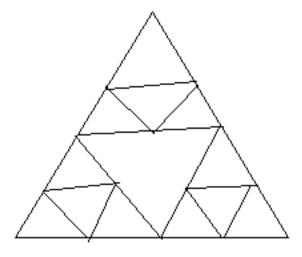
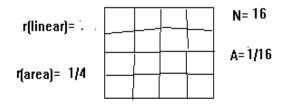


Figure 4: Parallel . Sierpinski gasket



D=log N/ log (1/r) = log 16/ log 1/16 log 16/log 4=2

Figure 5: Parallel fractals

IV. Computer Generating Fractals

Computer generation fractals are any type of graphics with sale-similarities [3]. The recursion technical is used to generate fractals with self-similarities. The applications of fractal graphics ranging from graphics design to designing fractals on garments Python Programming is to generate Fractal Graphics [2]

```
# Fractal generation with phython
from turtle import *
def mink(lengthSide, levels):
       if levels == 0:
               forward(lengthSide)
               return
       lengthSide /= 3.0
       mink(lengthSide, levels-1)
       left(60)
       mink(lengthSide, levels-1)
       riaht(120)
       mink(lengthSide, levels-1)
       left(60)
       mink(lengthSide, levels-1)
if __name__ == "__main__":
       speed(0)
```

speed(0)
length = 300.0
penup()
backward(length/2.0)
pendown()
for i in range(4):
mink(length,4)
right(120)
mainloop()

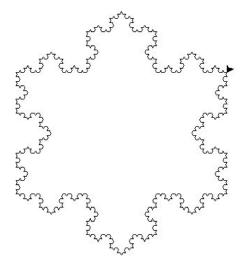


Figure 6: Koch curve

V. Conclusion

Fractals are structures which having the property of Scale-invariance or self-similarity. Self-similarity of a system implies that features of a structures are lookalike similar structures at different scales of length. Fractal Graphics have the number of applications in designing clothes and crafts. These a pplications are describes as fractals. Fractal dimension will identify the fractal structures or not. For instance, circles are not fractals. Different fractal structures are studied. The methods and techniques are also

proposed for Computer generation of the fractals to simplify the process.

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Feature Matching with Improved SIRB using RANSAC

By Meet Palod, Manas Joshi, Amber Jain, Viroang Rawat & Prof. K.K. Sharma

Abstract- In this paper, we suggest to improve the SIRB (SIFT (Scale-Invariant Feature Transform) and ORB (Oriented FAST and Rotated BRIEF)) algorithm by incorporating RANSAC to enhance the matching performance. We use multi-scale space to extract the features which are impervious to scale, rotation, and affine variations. Then the SIFT algorithm generates feature points and passes the interest points to the ORB algorithm. The ORB algorithm generates an ORB descriptor where Hamming distance matches the feature points. We propose to use RANSAC (Random Sample Consensus) to cut down on both the inliers in the form of noise and outliers drastically, to cut down on the computational time taken by the algorithm. This post-processing step removes redundant key points and noises. This computationally effective and accurate algorithm can also be used in handheld devices where their limited GPU acceleration is not able to compensate for the computationally expensive algorithms like SIFT and SURF. Experimental results advocate that the proposed algorithm achieves good matching, improves efficiency, and makes the feature point matching more accurate with scale in-variance taken into consideration.

GJCST-F Classification: G.4



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Feature Matching with Improved SIRB using **RANSAC**

Meet Palod α. Manas Joshi σ. Amber Jain ρ. Viroang Rawat ω & Prof. K.K. Sharma ¥

Abstract- In this paper, we suggest to improve the SIRB (SIFT (Scale-Invariant Feature Transform) and ORB (Oriented FAST and Rotated BRIEF)) algorithm by incorporating RANSAC to enhance the matching performance. We use multi-scale space to extract the features which are impervious to scale, rotation, and affine variations. Then the SIFT algorithm generates feature points and passes the interest points to the ORB algorithm. The ORB algorithm generates an ORB descriptor where Hamming distance matches the feature points. We propose to use RANSAC (Random Sample Consensus) to cut down on both the inliers in the form of noise and outliers drastically, to cut down on the computational time taken by the algorithm. This post-processing step removes redundant key points and noises. This computationally effective and accurate algorithm can also be used in handheld devices where their limited GPU acceleration is not able to compensate for the computationally expensive algorithms like SIFT and SURF. Experimental results advocate that the proposed algorithm achieves good matching, improves efficiency, and makes the feature point matching more accurate with scale in-variance taken into consideration.

I. Introduction

eature extraction and the matching continue to be the cornerstone of computer vision technologies. Essentially in feature detection, we abstract the image information into numbers and make a local decision at every point to see if there exists an image feature at that point. These local decisions taken at multiple spots are what we call interest points or key points. Ideally, techniques used for key point detection and matching should be impervious to different changes like - rotation, scale variance, illumination changes, noise and perspective changes.

Another desirable characteristic of these algorithms is that the key points generated should be unique to a greater degree to match a single feature with high rates of success.

The process of feature detection and matching can be broken down into some essential steps:

- Detecting Interest points It is done based on the brightness of the image or on the boundary extraction method.
- 2. Description of interest points We generate a description vector for each feature point that describes the local appearance around every

- feature point that is invariant under changes in illumination, translation, scale, and in-plane rotation.
- Matching the interest points across images We match similar features across the images and map them, establishing a connection between two similar images.

Now, these general steps can be implemented using distinct Algorithms. To this extent, numerous algorithms have been proposed. The SIFT algorithm has been historically the most popular algorithm due to its application in several fields using visual features like object recognition, image stitching, visual mapping etc. of computational burden overhead accompanies SIFT led to an intensive search for its replacement.

Real-time systems and low-power devices like cell phones cannot bear that computational burden. SIFT generally performs better than SURF computational cost and is the most accurate featuredetector-descriptor for scale, rotation and affine variations.

SURF detects more features than the SIFT algorithm and the features are detected in a scattered form generally all over the image. SURF has the highest computational cost for feature matching. Using SURF on low-end devices was not possible, and so a faster and more efficient algorithm was required.

ORB detects the highest number of features that are more concentrated on corners. In all practical fields of comparison ORB performs better than SURF and SIFT or any other algorithm for that matter. It is computationally the fastest and most efficient.

SIRB is the latest addition to this arsenal of keypoint detection and matching algorithms. The SIRB algorithm conflates the accuracy of the SIFT algorithm and combines it with the computationally efficient algorithm of ORB, thus making it fast as well as accurate while also being unaffected by scale, rotation and affine variations. We have proposed to take this one step further and improve on the existing SIRB(SIFT+ORB) algorithm using RANSAC(Random sample consensus). The idea behind using RANSAC is to reduce the number of interest points and thus eliminating the non matching points. This can potentially reduce more than 50% of the points and because the number of points to be matched have been reduced we can considerably reduce the matching time.

The RANSAC algorithm removes outliers or the data which does not fit the model. Its application in feature matching comes from the fact that it can handle more than 50% of the data as outliers, if applicable, and that has a drastic effect on time needed to generate a result. It is worth mentioning that RANSAC is a non-deterministic algorithm and that the results produced are reasonable only with a certain probability. This probability increases with the number of iterations. In RANSAC, we broadly perform these two steps iteratively:

- Select a Minimal Sample Set(MSS) randomly from the input dataset and generate a model based on only these elements.
- We now test which elements of the entire dataset are consistent with the model generated using the MSS. The set of elements which fit the model constitute a consensus set.

This process terminates when the probability of finding a better consensus set drops below a certain threshold.

The proposed method will reduce the number of points to be matched and offer a better alternative to the existing SIRB algorithm.

A feature or a keypoint is the distinctive piece of information, which is used for image matching, image stitching, and image registration. In an image, features are the specific structures like edges, corners, or objects. We detect corner points from an image which are also called interest points. An interest point in an image has a well defined position. These features extracted through the SIFT algorithm are invariant to rotation, scaling, and partially to illumination.

Keypoints are important because no matter how the image changes, we find the same key points in the modified image when comparing with the original image.

To detect keypoints, SIFT starts by generating the scale space for the image. Image at different scales is blurred by convolving a Gaussian kernel.

Scale-space is divided into octaves. In the original paper, Lowe [1] suggested that 4 octaves and 5 blur levels at each octave are ideal for the algorithm.

Each octave's image size is half of the previous octave and each image within an octave is increasingly blurred by a factor of k. The image is progressively blurred within an octave.

Mathematically, blurring is convolving the Gaussian operator and the image.

$$L(x, y, \sigma) = G(x, y, \sigma) * I(x, y)$$
(1)

Here, G is the Gaussian kernel, I is an image, x, y are the location coordinates and defines the amount of blur. A bigger value of implies more blur. Now with the blurred images, we create another set of images, known as Difference of Gaussians (DoG). This DoG images are

used for finding keypoints in the image. It is obtained as the difference of two progressively blurred images in an octave. This process is done for all octaves of the image in the scale space, and the desired scale space is obtained. It is represented in below image:

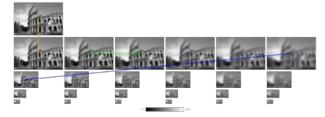


Fig. 1: Gaussian pyramid and Difference-of-Gaussian (DoG) pyramid

II. LOCAL MAXIMA/MINIMA DETECTION

For computing local maxima or minima in the Difference of Gaussian, each point is compared to its eight neighbors in the image, 9 in the scale above and 9 in the scale below. Pixel is considered as a keypoint if its pixel intensity is more than all of the 26 neighboring points. Also, it is computationally efficient, because points are eliminated in the first few inspections.

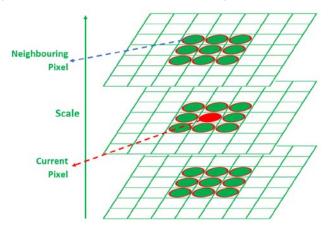


Fig. 2: Maxima and Minima are computed around the 3-dimensions x, y and σ .[15]

III. KEY POINT LOCALIZATION REMOVING KEYPOINTS WITH LOW CONTRAST

After finding a possible keypoint by comparing it to neighboring pixels, the next task is to find datapoint location, scale, and ratio of principal curvatures. This process allows us to reject data points with low contrast which is sensitive to noise or localized along an edge. After it we fit a 3D quadratic function to the sample points to resolve the interpolated location of maximum and minimum. This approach uses Taylor expansion (2nd order) of the scale-space function, D (x, y, σ).

$$D(x) = D + \frac{\partial D^{T}}{\partial x}x + \frac{1}{2}x^{T}\frac{\partial^{2}D}{\partial x^{2}}x \tag{2}$$

Where D and the derivatives of D are calculated at the keypoint and $x=(x, y, \sigma)$ T is the o set from this point. The correct location of keypoint extrema is x, and is evaluated by calculating the derivative of function D(x) concerning x and placing it to zero.

$$\hat{x} = -\frac{\partial^2 D^- 1}{\partial x^2} \frac{\partial D}{\partial x} \tag{3}$$

The function value at extrema, (i.e. at \hat{x}) D (\hat{x}) is useful for removing keypoints with low contrast. The value D (\hat{x}) can be calculated by substituting value of bx in $D(\widehat{x})$.

$$D(\hat{x}) = D + \frac{1}{2} \frac{\partial D^T}{\partial x} x \tag{4}$$

Values less than 0.3 for the function value D(\hat{x}) are rejected. A final trial for removing key points/feature points on edges are performed because these are unnecessary points and will create redundant key points in the matching process. A poorly defined extrema located at the ridge in the DoG, which depicts the edge in the image has broader principle curvature across the ridge and a low one along with it which is not present in a blob or (corner) having broad principle curvature along with both directions. Principle of curvature is evaluated by Hessian Matrix H.

$$H = \begin{bmatrix} D_{xx} & D_{xy} \\ D_{xy} & D_{yy} \end{bmatrix} \tag{5}$$

The eigenvalues corresponding to the Hessian matrix are tantamount to principle curvature. The ratio of eigenvalues $\lambda 1$ and $\lambda 2$ (which is proportional to principle curvature), are calculated with the help of Hessian matrix (H). This ratio is compared to certain threshold ratio r and high ratio points are repudiated. Where $r = \lambda 1/\lambda 2$.

$$Tr(H) = D_{xx} + D_{yy} = \lambda_1 + \lambda_2 \tag{6}$$

$$Det(H) = D_{xx}D_{yy} - (D_{xx})^2 = \lambda_1\lambda_2 \tag{7}$$

$$\frac{Tr(H)^2}{Det(H)} = \frac{(\alpha+\beta)^2}{\alpha\beta} = \frac{(r\beta+\beta)^2}{r\beta^2} = \frac{(r+1)^2}{r}$$
(8)

Apply homogeneity to equations. The value of the term $(r+1)^2/r$ is minimum when both eigenvalues are equal, and it gradually increases with the value of r. Hence to check whether the ratio is below certain value, we used

$$\frac{Tr(H)^2}{Det(H)} = \frac{(\lambda_1 + \lambda_2)^2}{\lambda_1 \lambda_2} < \frac{(r+1)^2}{r}$$
 (9)

We use the value of r = 10 in this paper. Feature points above this threshold are repudiated.

Orientation Assignment IV.

Assigning orientation to keypoints ensures that it can be represented relative to its orientation and thus rotation invariance is achieved. To assign orientation, Histogram of Oriented Gradient (HOG) is used.

The scale of the keypoint is used to select the Gaussian smoothed image. Then, a 16x16 square window is considered around the detected keypoint.

For each pixel, L (x, y), at this scale, the gradient magnitude, m (x, y), and edge orientation, (x, y), is precomputed using pixel differences:

$$\theta(x,y) = tan^{-1} \left(\frac{L(x,y+1) - L(x,y-1)}{L(x+1,y) - L(x-1,y)} \right)$$
(10)

The edges below a threshold gradient magnitude are considered as weak edges and discarded. From the surviving edges, an orientation histogram is created. It has 36 bins covering the 360degree range of orientations. Each pixel orientation added to the histogram is weighted by its gradient magnitude and by a Gaussian-weighted circular window with a that is 1.5 times that of the scale of the keypoint.

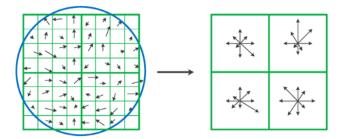


Fig. 3: Histogram of Gradient(HOG)[14]

Peaks in HOG correspond to the dominant orientation of the keypoint. If other peaks are detected within 80 percent of the dominant peak, then multiple keypoints are generated at the same location and scale with different orientation. These multiple orientation keypoint notably enhance the stability of matching.

rBRIEF: ROTATION-AWARE BRIEFS

After detecting all the key points, descriptors are generated for each keypoint. Feature descriptors encode unique information into a series of numbers and act like a numerical "fingerprint" that can be used to differentiate one feature from another. In this paper, we use the rBRIEF algorithm, a rotation-invariant version of the BRIEF algorithm. The BRIEF descriptor was proposed by M. Calonder [9]. BRIEF is an acronym for Binary robust independent elementary features, and it is robust to photometric and geometric transformation.

The defined neighborhood, which is a square of some pixel width and height, around the keypoint is known as a patch. Brief depends on intensity difference tests to represent an image patch as a binary vector. Image patches could be effectively classified based on small number of pairwise intensity a relatively comparisons.

Each keypoint is described by a feature vector. Brief convert image patches into binary feature vectors, which are 128-512 bits string, so that together they can represent an object. The binary feature vector contains only 1's and 0's. Because BRIEF deals at the pixel level, it is very noise-sensitive. Hence it starts by smoothing the image with a Gaussian kernel. This reduces the sensitivity and also increases the stability and repeatability of the descriptors by pre-smoothing the patch.

To generate a 256-bit vector, it defines a randomly generated set of 256 pairs of pixels. The first pair in the random pair is taken from a Gaussian distribution drew around the keypoint with a standard deviation of 0.04 * S² (where S is the dimension of the patch). The second pair in the random pair is drawn from a Gaussian distribution drawn around the first pixel(x) with a standard deviation of 0.01 * S². rBRIEF achieves rotation invariance by moving these pixel locations by an angle equal to the orientation angle of the keypoint.

Then, it performs a binary test comparing the intensity of the first pixel in pair with the second pixel. If the intensity of the first pixel is less than the intensity of the second pixel, we append value 1 in the feature vector, else we append 0. When this binary test is performed on all the 256-pixel pairs in an image patch, we get a 256-bit binary vector, which is the desired descriptor.

VI. RANSAC

RANSAC[8] algorithm is an extensively adapted mechanism in image processing for cleaning outliers from huge datasets. This algorithm helps in outlier removal, image noise filtration and optimal descriptors selection. RANSAC[6][7] does statistical estimations, which helps in evaluation for the likelihood of achieving accurate predictions.

RANSAC uses a minimal amount of data set to gain noise-repudiated data points, unlike other algorithms which initiate with huge datasets and then remove outliers.

Working of RANSAC:

- First, a minimum number of points are arbitrarily selected to define model parameters. From all the points, the points which fit according to the tolerance €(epsilon) are determined.
- Then we calculate the fraction of the number of inliers to total points present in the set.
- If the fraction is greater than a threshold (tau), which is predefined, we reevaluate parameters of the model using inliers.
- Else, we reiterate from step 1 to step 4, at most N times.

In this procedure, the value of N is taken large enough so that the minimum of one of the sets of random samples does not include any outlier. P represents probability which is set to 0.99, U represents chances of a selected data point as an inlier, V represents chances of selected data point as an outlier and N times the minimum number of points denoted M are required, so

$$1 - P = (1 - U^M)^N (11)$$

After some formulations, N is chosen by the formula:

$$N = \frac{\log(1-p)}{\log(1-(1-V)^M)}$$
 (12)

In our methodology, we aim to improve performance of SIRB by using RANSAC. RANSAC has been proven highly proficient in removing mismatched inconsistent sets of points. Since SIRB incorporates SIFT and ORB, therefore RANSAC has an intrinsic role in boosting the overall performance of SIRB. Due to the inaccuracy of SIFT, some points generated by it are mapped incorrectly, but RANSAC helps in filtering out these points, making SIFT more eficient. ORB [5] is an essential feature descriptor in low-end devices. So, it becomes crucial to eliminate shortcomings of ORB [12] and hence RANSAC is used as a post-quality improvement step to remove outliers and redundant key-points. RANSAC, thus makes SIRB more efficacious than some pre-existing methods, consequently making it an overall vigorous image recognition set-up.

VII. Match Feature Points Via Hamming Distance

The binary feature descriptor generated makes the matching process computationally more efficient because these binary vectors are stored in the form of bits. Performing operations like XOR on these binary vectors are much faster because of the ability to quickly compare descriptor pairs using few processor-level instructions. Algorithms like SIFT/SURF use Euclidean distance to compare binary feature vectors. We found that using hamming distance[11] for matching the key points is more efficient when comparing binary feature vectors. Therefore, we have used Hamming distance for our research.

The hamming distance is the value of the number of positions where both binary vectors differ. It is denoted by d(a, b), where a and b are two binary vectors. It can also be calculated by performing XOR operation on given two binary vectors.

$$d_{hamming}(a,b) = \sum_{i=0}^{n-1} (a_i \oplus b_i)$$
 (13)

VIII. Conclusion



Fig. 4: Matching Result for Images with Scale Changes

We proposed a new algorithm, SIRB+RANSAC. It inherits the innate accuracy of the SIFT algorithm and the fast superiority of ORB. Its performance is further enhanced by using the RANSAC algorithm, which eliminates mismatches to a great extent. We start by describing the SIFT keypoint detection algorithm. We then introduced the rBrief method for binary descriptor generation. We used Hamming Distance to match the binary descriptors generated. Finally, we used the RANSAC algorithm to remove the mismatched keypoint pairs and eliminate the outliers.

The results of the experiment show that the new enhanced SIRB+RANSAC algorithm effectively resolves the problem that the ORB algorithm performs poorly in terms of scale invariance, and it is sensitive to image illumination. Based on the results of the experiment, we conclude that:

- The average matching accuracy of our algorithm is 89.85% as compared to the 86.95% accuracy of the traditional SIRB algorithm. The SIFT algorithm still presents an accuracy of 94.20%.
- The average computational time of the new improved SIRB+RANSAC algorithm is 87.69ms, which is about 42 times faster than the SIFT algorithm which takes an average of 3723ms. On the other hand, the SIRB algorithm takes 74.33ms.

The proposed algorithm will be exceptionally useful on low-end hand-held devices with low GPU power because it involves relatively less computational expenses than the prevalent algorithms. It also has applications in traditional systems where SIFT or SURF is currently being used.

FUTURE WORK IX.

It is evident from experimental results that the average matching time for SIRB is significantly less when compared with the SIFT algorithm. However, the matching accuracy has also dropped by 4.61%. We have released the source code of our algorithm so that it can be used by other researchers. There is the scope for improvement in our proposed algorithm, and these are the fields where future research can be directed:

Instead of using Gaussian filters for convolution, filters like average filter and median filters can be

- used to reduce the run time of the algorithm while maintaining accuracy.
- Use improved RANSAC, modified RANSAC or optimal RANSAC to remove parts of the error feature point, thus increasing the proportion of correct matching features.
- Remove the repeatable and unstable key points generated after normalizing the scale space.

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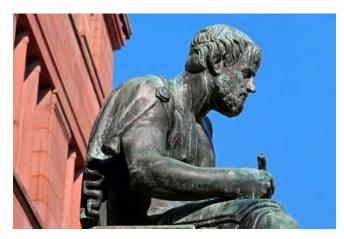
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- 3. Final approval of the version of the paper to be published.

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Unless specified in the notification, the Editorial Board's decision on publication of the paper is final and cannot be appealed before making the major change in the manuscript.

Acknowledgments

Contributors to the research other than authors credited should be mentioned in Acknowledgments. The source of funding for the research can be included. Suppliers of resources may be mentioned along with their addresses.

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The following is the official style and template developed for publication of a research paper. Authors are not required to follow this style during the submission of the paper. It is just for reference purposes.



Manuscript Style Instruction (Optional)

- Microsoft Word Document Setting Instructions.
- Font type of all text should be Swis721 Lt BT.
- Page size: 8.27" x 11'", left margin: 0.65, right margin: 0.65, bottom margin: 0.75.
- Paper title should be in one column of font size 24.
- Author name in font size of 11 in one column.
- Abstract: font size 9 with the word "Abstract" in bold italics.
- Main text: font size 10 with two justified columns.
- Two columns with equal column width of 3.38 and spacing of 0.2.
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- The paragraph before spacing of 1 pt and after of 0 pt.
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- Large images must be in one column.
- The names of first main headings (Heading 1) must be in Roman font, capital letters, and font size of 10.
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Structure and Format of Manuscript

The recommended size of an original research paper is under 15,000 words and review papers under 7,000 words. Research articles should be less than 10,000 words. Research papers are usually longer than review papers. Review papers are reports of significant research (typically less than 7,000 words, including tables, figures, and references)

A research paper must include:

- a) A title which should be relevant to the theme of the paper.
- b) A summary, known as an abstract (less than 150 words), containing the major results and conclusions.
- c) Up to 10 keywords that precisely identify the paper's subject, purpose, and focus.
- d) An introduction, giving fundamental background objectives.
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- Results which should be presented concisely by well-designed tables and figures.
- g) Suitable statistical data should also be given.
- h) All data must have been gathered with attention to numerical detail in the planning stage.

Design has been recognized to be essential to experiments for a considerable time, and the editor has decided that any paper that appears not to have adequate numerical treatments of the data will be returned unrefereed.

- i) Discussion should cover implications and consequences and not just recapitulate the results; conclusions should also be summarized.
- j) There should be brief acknowledgments.
- k) There ought to be references in the conventional format. Global Journals recommends APA format.

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Abstract

The abstract is the foundation of the research paper. It should be clear and concise and must contain the objective of the paper and inferences drawn. It is advised to not include big mathematical equations or complicated jargon.

Many researchers searching for information online will use search engines such as Google, Yahoo or others. By optimizing your paper for search engines, you will amplify the chance of someone finding it. In turn, this will make it more likely to be viewed and cited in further works. Global Journals has compiled these guidelines to facilitate you to maximize the webfriendliness of the most public part of your paper.

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Authors are advised to submit any mathematical equation using either MathJax, KaTeX, or LaTeX, or in a very high-quality image.

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Tables: Tables should be cautiously designed, uncrowned, and include only essential data. Each must have an Arabic number, e.g., Table 4, a self-explanatory caption, and be on a separate sheet. Authors must submit tables in an editable format and not as images. References to these tables (if any) must be mentioned accurately.



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TIPS FOR WRITING A GOOD QUALITY COMPUTER SCIENCE RESEARCH PAPER

Techniques for writing a good quality computer science research paper:

- 1. Choosing the topic: In most cases, the topic is selected by the interests of the author, but it can also be suggested by the guides. You can have several topics, and then judge which you are most comfortable with. This may be done by asking several questions of yourself, like "Will I be able to carry out a search in this area? Will I find all necessary resources to accomplish the search? Will I be able to find all information in this field area?" If the answer to this type of question is "yes," then you ought to choose that topic. In most cases, you may have to conduct surveys and visit several places. Also, you might have to do a lot of work to find all the rises and falls of the various data on that subject. Sometimes, detailed information plays a vital role, instead of short information. Evaluators are human: The first thing to remember is that evaluators are also human beings. They are not only meant for rejecting a paper. They are here to evaluate your paper. So present your best aspect.
- 2. Think like evaluators: If you are in confusion or getting demotivated because your paper may not be accepted by the evaluators, then think, and try to evaluate your paper like an evaluator. Try to understand what an evaluator wants in your research paper, and you will automatically have your answer. Make blueprints of paper: The outline is the plan or framework that will help you to arrange your thoughts. It will make your paper logical. But remember that all points of your outline must be related to the topic you have chosen.
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- 19. Refresh your mind after intervals: Try to give your mind a rest by listening to soft music or sleeping in intervals. This will also improve your memory. Acquire colleagues: Always try to acquire colleagues. No matter how sharp you are, if you acquire colleagues, they can give you ideas which will be helpful to your research.

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INFORMAL GUIDELINES OF RESEARCH PAPER WRITING

Key points to remember:

- Submit all work in its final form.
- Write your paper in the form which is presented in the guidelines using the template.
- Please note the criteria peer reviewers will use for grading the final paper.

Final points:

One purpose of organizing a research paper is to let people interpret your efforts selectively. The journal requires the following sections, submitted in the order listed, with each section starting on a new page:

The introduction: This will be compiled from reference matter and reflect the design processes or outline of basis that directed you to make a study. As you carry out the process of study, the method and process section will be constructed like that. The results segment will show related statistics in nearly sequential order and direct reviewers to similar intellectual paths throughout the data that you gathered to carry out your study.

The discussion section:

This will provide understanding of the data and projections as to the implications of the results. The use of good quality references throughout the paper will give the effort trustworthiness by representing an alertness to prior workings.

Writing a research paper is not an easy job, no matter how trouble-free the actual research or concept. Practice, excellent preparation, and controlled record-keeping are the only means to make straightforward progression.

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To make a paper clear: Adhere to recommended page limits.



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Mistakes to avoid:

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- Separating a table, chart, or figure—confine each to a single page.
- Submitting a manuscript with pages out of sequence.
- In every section of your document, use standard writing style, including articles ("a" and "the").
- Keep paying attention to the topic of the paper.
- Use paragraphs to split each significant point (excluding the abstract).
- Align the primary line of each section.
- Present your points in sound order.
- Use present tense to report well-accepted matters.
- Use past tense to describe specific results.
- Do not use familiar wording; don't address the reviewer directly. Don't use slang or superlatives.
- Avoid use of extra pictures—include only those figures essential to presenting results.

Title page:

Choose a revealing title. It should be short and include the name(s) and address(es) of all authors. It should not have acronyms or abbreviations or exceed two printed lines.

Abstract: This summary should be two hundred words or less. It should clearly and briefly explain the key findings reported in the manuscript and must have precise statistics. It should not have acronyms or abbreviations. It should be logical in itself. Do not cite references at this point.

An abstract is a brief, distinct paragraph summary of finished work or work in development. In a minute or less, a reviewer can be taught the foundation behind the study, common approaches to the problem, relevant results, and significant conclusions or new questions.

Write your summary when your paper is completed because how can you write the summary of anything which is not yet written? Wealth of terminology is very essential in abstract. Use comprehensive sentences, and do not sacrifice readability for brevity; you can maintain it succinctly by phrasing sentences so that they provide more than a lone rationale. The author can at this moment go straight to shortening the outcome. Sum up the study with the subsequent elements in any summary. Try to limit the initial two items to no more than one line each.

Reason for writing the article—theory, overall issue, purpose.

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- To-the-point depiction of the research.
- Consequences, including definite statistics—if the consequences are quantitative in nature, account for this; results of any numerical analysis should be reported. Significant conclusions or questions that emerge from the research.

Approach:

- Single section and succinct.
- An outline of the job done is always written in past tense.
- o Concentrate on shortening results—limit background information to a verdict or two.
- Exact spelling, clarity of sentences and phrases, and appropriate reporting of quantities (proper units, important statistics) are just as significant in an abstract as they are anywhere else.

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The introduction should "introduce" the manuscript. The reviewer should be presented with sufficient background information to be capable of comprehending and calculating the purpose of your study without having to refer to other works. The basis for the study should be offered. Give the most important references, but avoid making a comprehensive appraisal of the topic. Describe the problem visibly. If the problem is not acknowledged in a logical, reasonable way, the reviewer will give no attention to your results. Speak in common terms about techniques used to explain the problem, if needed, but do not present any particulars about the protocols here.



The following approach can create a valuable beginning:

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- Present a justification. State your particular theory(-ies) or aim(s), and describe the logic that led you to choose them.
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Use past tense except for when referring to recognized facts. After all, the manuscript will be submitted after the entire job is done. Sort out your thoughts; manufacture one key point for every section. If you make the four points listed above, you will need at least four paragraphs. Present surrounding information only when it is necessary to support a situation. The reviewer does not desire to read everything you know about a topic. Shape the theory specifically—do not take a broad view

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Materials:

Materials may be reported in part of a section or else they may be recognized along with your measures.

Methods:

- o Report the method and not the particulars of each process that engaged the same methodology.
- Describe the method entirely.
- o To be succinct, present methods under headings dedicated to specific dealings or groups of measures.
- Simplify—detail how procedures were completed, not how they were performed on a particular day.
- o If well-known procedures were used, account for the procedure by name, possibly with a reference, and that's all.

Approach:

It is embarrassing to use vigorous voice when documenting methods without using first person, which would focus the reviewer's interest on the researcher rather than the job. As a result, when writing up the methods, most authors use third person passive voice.

Use standard style in this and every other part of the paper—avoid familiar lists, and use full sentences.

What to keep away from:

- o Resources and methods are not a set of information.
- Skip all descriptive information and surroundings—save it for the argument.
- Leave out information that is immaterial to a third party.



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The principle of a results segment is to present and demonstrate your conclusion. Create this part as entirely objective details of the outcome, and save all understanding for the discussion.

The page length of this segment is set by the sum and types of data to be reported. Use statistics and tables, if suitable, to present consequences most efficiently.

You must clearly differentiate material which would usually be incorporated in a study editorial from any unprocessed data or additional appendix matter that would not be available. In fact, such matters should not be submitted at all except if requested by the instructor.

Content:

- Sum up your conclusions in text and demonstrate them, if suitable, with figures and tables.
- o In the manuscript, explain each of your consequences, and point the reader to remarks that are most appropriate.
- o Present a background, such as by describing the question that was addressed by creation of an exacting study.
- Explain results of control experiments and give remarks that are not accessible in a prescribed figure or table, if appropriate.
- Examine your data, then prepare the analyzed (transformed) data in the form of a figure (graph), table, or manuscript.

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- Do not discuss or infer your outcome, report surrounding information, or try to explain anything.
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- o A manuscript should complement any figures or tables, not duplicate information.
- o Never confuse figures with tables—there is a difference.

Approach:

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Put figures and tables, appropriately numbered, in order at the end of the report.

If you desire, you may place your figures and tables properly within the text of your results section.

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Infer your data in the conversation in suitable depth. This means that when you clarify an observable fact, you must explain mechanisms that may account for the observation. If your results vary from your prospect, make clear why that may have happened. If your results agree, then explain the theory that the proof supported. It is never suitable to just state that the data approved the prospect, and let it drop at that. Make a decision as to whether each premise is supported or discarded or if you cannot make a conclusion with assurance. Do not just dismiss a study or part of a study as "uncertain."



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- o Give details of all of your remarks as much as possible, focusing on mechanisms.
- Make a decision as to whether the tentative design sufficiently addressed the theory and whether or not it was correctly restricted. Try to present substitute explanations if they are sensible alternatives.
- One piece of research will not counter an overall question, so maintain the large picture in mind. Where do you go next? The best studies unlock new avenues of study. What questions remain?
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Approach:

When you refer to information, differentiate data generated by your own studies from other available information. Present work done by specific persons (including you) in past tense.

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Topics	Grades		
	А-В	C-D	E-F
Abstract	Clear and concise with appropriate content, Correct format. 200 words or below	Unclear summary and no specific data, Incorrect form Above 200 words	No specific data with ambiguous information Above 250 words
Introduction	Containing all background details with clear goal and appropriate details, flow specification, no grammar and spelling mistake, well organized sentence and paragraph, reference cited	Unclear and confusing data, appropriate format, grammar and spelling errors with unorganized matter	Out of place depth and content, hazy format
Methods and Procedures	Clear and to the point with well arranged paragraph, precision and accuracy of facts and figures, well organized subheads	Difficult to comprehend with embarrassed text, too much explanation but completed	Incorrect and unorganized structure with hazy meaning
Result	Well organized, Clear and specific, Correct units with precision, correct data, well structuring of paragraph, no grammar and spelling mistake	Complete and embarrassed text, difficult to comprehend	Irregular format with wrong facts and figures
Discussion	Well organized, meaningful specification, sound conclusion, logical and concise explanation, highly structured paragraph reference cited	Wordy, unclear conclusion, spurious	Conclusion is not cited, unorganized, difficult to comprehend
References	Complete and correct format, well organized	Beside the point, Incomplete	Wrong format and structuring

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