Research Article

Adaptive Thresholding based Image Binarization Using VHDL

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Abstract: The aim of this study is to recognize the given image with the existing image based on the technique of image binarization by MATLAB tool and it is simulated using VHDL (Very High Speed IC Hardware Descriptive Language) using MODELSIM tool. This image binarization is based on LEGION (Locally Excitatory Globally inhibitory Oscillatory Network) Concept. In this study the algorithm finds an optimum threshold technique, the other by separating the image background and foreground pixels. This algorithm has superior performance in separating the images from background in comparison with the other threshold techniques.

Keywords: Adaptive threshold, binarisation, MATLAB, MODELSIM, VHDL

INTRODUCTION

To extract the required image from a given image, it may be separated into different components say foreground and background. An adaptive threshold concerning extraction of image from the background in a given image sequences is possible for implementation in hardware. The conventional histogram based threshold methods are deficient in detecting images due to poor contrasts between image and the background, or the change of illuminations (Otsu, 1979; Rosenfeld and De La Torre, 1983; Sezan, 1985; Trier and Taxt, 1995). But in real time applications, except adaptive threshold techniques are not fast enough for hardware implementation (Kittler and Illingworth, 1986). The image binarization (Lee et al., 1991; White and Rohrer, 1983) is a process of dividing the original image into two components, foreground or image and background. The gray level value of background (value ‘0’) is different from the foreground (value ‘255’) values. So, for the effective separation of foreground and background, this thresholding techniques may be considered (Hertz and Schafer, 1988; Sezign and Sankar, 2004). The thresholding techniques yields a binary image as an output by assigning pixels with values less than the threshold as 0’s and remaining pixels as 1’s. In this study the Image binarization is carried out by MATLAB tool and it is simulated using VHDL (Very High Speed IC Hardware Descriptive Language) using MODELSIM tool (XESS Corporation, 2001). This image binarization is based on LEGION (Locally Excitatory Globally inhibitory Oscillatory Network) Concept (Ligon et al., 1998).

The LEGION concept has the following features:

- Binarization is realized in a very short time due to the simple processing.
- Due to the simple structure of each cell representing a pixel compact integration of many cells on a single chip becomes possible.
- Fast software implementation is possible.

Neural oscillator concept: There are three possible ways to reach synchrony as shown in Fig. 1 to 3.

- LEGION is composed of a Basic Oscillator, Local Excitatory Connections (produce phase synchrony within each object) and Global Inhibitor (it receives inputs from the entire network and feeds back with
inhibition). Each Oscillator will respond to a detected feature at some location in the image. Synchronized group of oscillators is separated from other synchronized groups that represent different objects by de-synchronization. It is proposed in Digital implementation, Model described by digital algorithm, both gray/color scale binarization and segmentation in a single algorithm (Yanowitz and Bruckstein, 1989). Reprogrammability is possible using FPGA (Shirazi et al., 1995).

The Fig. 4 shows Digital LEGION Image Binarization Architecture. It consists of six blocks namely:

- Input ROM
- Weight calculation
- Leader cell
- Locally excitatory
- Segregation/Inhibition
- Output RAM

**DIGITAL LEGION ALGORITHM**

**Initialization:**

**Step 1:** Initial global inhibitor, \( Z = 0 \)

**Step 2:** Calculation of Weights

**Gray scale images:**

\[
W_{ij} = \frac{I_{\text{max}}}{1 + \delta |I_i - |j|} 
\]

where,

\( W_{ij} \) = Weight value between \( i^{th} \) and \( j^{th} \) pixels
\( I_i \) = \( i^{th} \) pixel value
\( I_j \) = \( j^{th} \) pixel value
\( I_{\text{max}} \) = Max. Gray level value \( [2^n - 1] \)

**Color image:**

\[
W_{(R)ij} = \frac{I_{(R)\text{max}}}{1 + |I_{(R)} - |j|} \\
W_{(G)ij} = \frac{I_{(G)\text{max}}}{1 + |I_{(G)} - |j|} \\
W_{(B)ij} = \frac{I_{(B)\text{max}}}{1 + |I_{(B)} - |j|} \\
W_{ij} = \min \{W_{(R)ij}, W_{(G)ij}, W_{(B)ij}\}
\]

**Step 3:** Detection of Leader Cell (P)

\[
\sum_{j=1}^{N} \text{If} \ ( \delta W_{ij} > Ti) \ \text{then} \ P_i = 1, \ \text{otherwise} \ P_i = 0
\]

where,

\( W_{ij} \) = weight value between \( i^{th} \) pixel and \( j \) neighboring pixels
\( j \) = 1, 2, 3…N
\( Ti \) = Leader cell threshold value

**Step 4:** Set all cell to non-excitation \( X_i(0) = 0 \).
Self-excitation of leader cell: If (excitable cells = false) then stop; //terminate Otherwise (find leader $P_i = 1$) then
$$X_i = 1, \ Z_i = 1; \ \text{goto (step 3) //Self-Excitation Otherwise (step2);}$$

Identification of dependent cell (or) local excitation:
If pixel $I_i$ is a neighbor of $X_i = 1$ and $W_iP_i > T_2$ then $I_i$ is a dependent cell of $X_i$.
Where, $W_iP_i$ = weight between pixel $I_i$ and weight corresponding to $X_i = 1$ location.

Segregation:
- Identify first excitation pixel $X_i = 1$ and read corresponding pixel value $I_i$
- Identify $X_i + 1 = 1$ pixel, read corresponding pixel value ($I_i + 1$)
- If $I_i + 1 - I_i < T_3$ then $X_i + 1 = 1$
Table 1: Input ROM

|    | 36 | 36 | 127 | 24 |
|----|----|----|-----|----|
| 24 | 24 | 26 | 36  | 127|
| 127| 24 | 24 | 36  | 36 |
| 128| 24 | 24 | 24  | 127|
| 36 | 24 | 127| 24  | 36 |

Table 2: Output of weight calculation block

|    | 276 | 550 | 786 | 263 | 23 |
|----|-----|-----|-----|-----|----|
| 276| 573 | 1060| 1060| 263 |
| 276| 573 | 826 | 826 | 533 |
| 344| 476 | 1043| 826 | 10 |
| 24 | 280 | 135 | 533 | 40 |

Table 3: Local threshold is 200. Output of a leader cell-pixel position

|    | 0   | 0   | 0   | 0   |
|----|-----|-----|-----|-----|
| 0  | 0   | 0   | 1   | 0   |
| 0  | 0   | 1   | 0   | 0   |
| 0  | 0   | 1   | 0   | 0   |
| 0  | 0   | 0   | 0   | 0   |

Table 4: Output of a local excite cell-position

|    | 1   | 1   | 1   | 0   |
|----|-----|-----|-----|-----|
| 0  | 1   | 1   | 1   | 0   |
| 0  | 1   | 1   | 1   | 1   |
| 0  | 1   | 1   | 1   | 0   |
| 0  | 1   | 0   | 1   |

- Keep I_i value as same for all other excitation pixel X_i+2, X_i+3 ... X_i+N
- Repeat the process for all X_i+N = 1 pixels; N = 1, 2, 3...

Table 5: Output of leader cell 2-pixel position segregation threshold is 10

|    | 1   | 0   | 0   | 0   |
|----|-----|-----|-----|-----|
| 0  | 1   | 0   | 0   |
| 0  | 0   | 1   |
| 0  | 0   | 1   | 1   |
| 0  | 1   |

Table 6: Output of segregation block-pixel position

|    | 2   | 0   | 0   | 0   |
|----|-----|-----|-----|-----|
| 0  | 2   | 1   | 1   |
| 0  | 0   | 2   | 0   |
| 0  | 0   | 2   | 2   |
| 0  | 2   |

Inhibit all segregated output: First segmentation is over, if still more segmentation is there goto step 3 and repeat the algorithm. The flow chart (Fig. 5) is the description of LEGION algorithm.

RESULTS AND DISCUSSION

The following (Fig. 6 and 7) are the sample image and its binarization.

The sample image is given to MATLAB Tool and the Corresponding Matrix values are obtained. The following tables (Table 1 to 6) shows the output of the (Input ROM, weight calculation, leader cell 1 and 2, local excite and output of RAM respectively) in the digital LEGION image binarization architecture.
The weight value between 36-24 is 19, 127-36 is 2, 127-24 is 2 and 127-128 is 127.
Let the leader threshold value be 1000.
The corresponding simulated output using VHDL is shown in Fig. 8.

CONCLUSION

In the present study, the image binarization is done by MATLAB and the same is simulated using VHDL by using MODELSIM. In this study the algorithm finds an optimum threshold technique, the other by separating the image background and foreground pixels. This algorithm has superior performance in separating the images from background in comparison with the other threshold techniques.

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