

Removing Spatial Redundancy from Image by Using Variable Vertex Chain Code

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Abstract:

In this paper we have proposed an efficient lossless image compression method, this method consists of two parts, the first is compressed part that based on two approaches of chain code representation the first approach is variable vertex chain code (V_VCC) is used to encode boundary of segment to remove spatial redundancy in image, the second approach is Freeman chain code 8-directional (FCCE) is used to encode remaining values then the values that belong to two approaches of chain code will be removed from image and reduce the size of image then begins a new level and will apply the same steps as the previous, the second is decompressed part , depending on variable vertex chain code set, Freeman chain code set, start points and values decompressed part will rebuild the original image identically.

Key words: Chain code, Lossless image compression, FCCE, Segmentation, V_VCC.

1. Introduction

Computer imaging plays an important role in many areas

ranging from consumer digital photo albums to remote earth sensing. The growing production of images and demands to their quality require high performance compression systems for efficient transmission, storage and archival (Krivoulets 2004).

Various types of redundancy exist in images, such as temporal redundancy, spatial redundancy (or interpixel redundancy), coding redundancy, spectral redundancy and psychovisual redundancy. The main goal of image compression is to minimize the number of bits required to represent the original images by reducing the redundancy in images, the essential issue in image compression is to design efficient and effective compression schemes (Zha 2007).

In terms of decompression ability, image compression systems can be generally classified into two main types: *lossless* image compression and *lossy* image compression.

Lossy compression involves some loss of information. Data that have been compressed using lossy techniques cannot be fully reconstructed. In return for accepting this loss, a higher compression ratio can be gained in comparison to lossless techniques. Generally, the applications of lossy compression are limited to the media domain (i.e., images, videos and audio (Kattan 2010).

If data have been lossless compressed, the original data can be reconstructed exactly from the compressed data. It is generally used for applications that cannot tolerate any difference between the original and reconstructed data (Panrong 2001)

Regardless of the compression type, all algorithms share a similar process when compressing information, namely: modelling and coding

The First stage, we recognize the sequence, look for regularities and similarities, this is done in the modeling stage, the modeling method is specialized for the type of data we compress, and it is clear that in video data we will be searching for different similarities than in text data. The modeling

systems are often different for lossless and lossy methods. Selecting the suitable modeling method is important because the more regularity we find the more we can reduce the sequence length. In particular, we cannot reduce the sequence length at all if we do not know what is redundant in it.

The second stage, coding, is based on the knowledge obtained in the modeling stage, and removes the redundant data, the coding systems are not so diverse because the modeling process is the stage where the adaptation to the data is made, therefore we only have to encode the sequence efficiently removing known redundancy (Deorowicz 2003).

Image compression can be achieved by *coding domain methods*, *spatial domain methods*, *transform domain methods* or a combination of these methods. Coding methods are directly applied to the raw images handling them as a sequence of discrete numbers, common coding methods include: Arithmetic, Huffman, Lempel-Ziv Welch (LZW), and Run-Length.

Spatial domain methods, which are a combination of spatial domain algorithms and coding methods, not only run directly on the gray values in an image but also try to remove the spatial redundancy, In transform domain compression; the image is represented using an suitable basis set, and the goal is to obtain a sparse coefficient matrix, Discrete Cosine transform (DCT) based compression and wavelet transform are two examples of transform domain methods. Recently, there has been a growing interest in reversible/lossless image compression of medical images (Vernuri et al. 2007).

2. Chain codes

Chain codes are used to represent a boundary by a connected sequence of straight-line segments of specified length and direction, typically this representation is based on 4- or 8-connectivity of the segments. The direction of each segment is coded by using a numbering system such as the ones shown in

Figure (1). Digital images usually are acquired and processed in a grid format with equal spacing in the x - and y -directions, so a chain code could be created by following a boundary in, say, a clockwise direction and assigning a direction to the segments connecting every pair of pixels. This method generally is unacceptable for two principal reasons: (1) The resulting chain of codes tends to be quite long, and (2) any small disturbances along the boundary due to noise or imperfect segmentation cause changes in the code that may not be related to the shape of the boundary (Gonzalez & R. E. Woods, 2002).

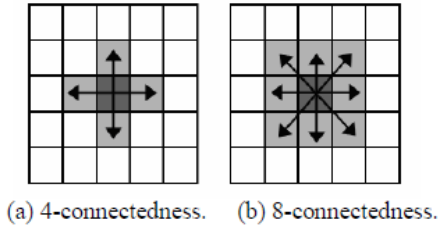


Figure (1) Different definitions of pixel neighborhoods

There are several types have been developed based on Freeman chain Code, briefly we'll show below the main types of chain code:

i. Original Vertex Chain Code (VCC): In 1999, E. Bribiesca first introduced the original vertex chain code (VCC). The VCC is based on the numbers of cell vertices which are in touch with the bounding contour of the shape. This code determines the number of pixels of the binary shape that are in touch with the observed vertex of the shape's boundary contour.

Figure (3 A) is an illustration of VCC which presents a shape. It is previous that there are only three numbers of 0, 1, and 2 which are needed to present a boundary composed with pixels of quadrate grids (Li et al. 2012).

ii. Variable-Length Vertex Chain Code (V_VCC) The second chain code that based on vertex chain code is variable-

length vertex chain code (V_VCC). This chain code also uses rectangular cells code of VCC. It has three elements 1, 2, and 3. These elements are represented by binary digits (0, 10, and 11). Variable-length VCC (V_VCC) code is well-defined by the probability of three code occurring. The experiment presented that the probability of code 2 is greater than the other two. Therefore, at V_VCC the binary digit 0 is used to represent code 2, 10 for code 1, and 11 for code 3 by applying the concept of Huffman coding, As depicted in figure (3 B) (Wulandhari and Haron 2008).

iii. Orthogonal chain code (3ot) the 3OT is based on the three relative changes of orthogonal directions given by Figure (3 C). In the 3OT, there are three elements: 0, 1 and 2. The element “0” represents no direction change which means to “go straight” through the contiguous straight line segments following the direction of the last segment; the “1” indicates a direction change forward with regard to the basis segment; and “2” means to “go back” with regard to the direction of the basis segment (Sanchez-Cruz 2010).

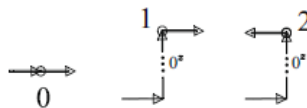


Figure (2) Visual meaning of the 3OT code

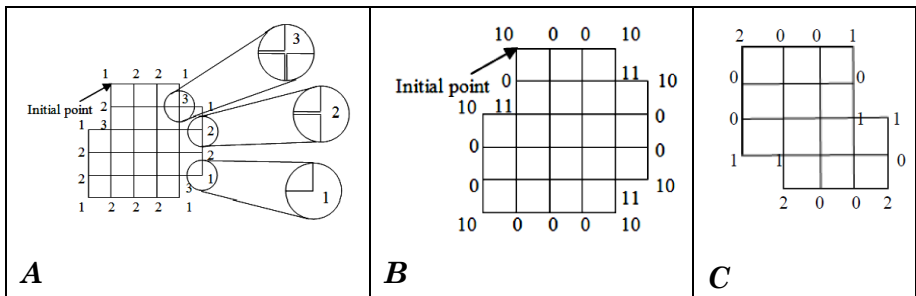


Figure (3) Visual meaning of the 3OT code

We will take an example then compare the methods (FCCE, 3OT, VCC, V_VCC) of chain code to select the best in term of efficiency of compression ratio which use it to encode boundaries, which produces from the segmentation process in proposed system.

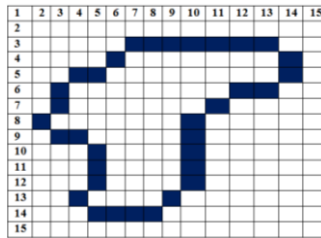


Figure (4) shape

The result of encoding the shape in figure (4) is illustrated in table (1).

Method	encoding	Number of symbols	Numbers of bits
FCCE	0,0,0,0,0,0,7,6,5,4,5,5,6,6,6,6,5,5,4,4,4,3,1,2,2,3,4,3,1,6,1,0,1,1	34	102
3OT	1,0,0,0,0,0,2,1,1,0,2,1,1,0,1,1,1,0,0,0,0,1,1,1,1,1,0,0,2,1,1,2,1,0,0,1,0,1,1,1,2,1,0,1,1,1,0,1,1,1,1	52	104
VCC	1,2,2,2,2,2,2,1,3,1,2,1,3,1,2,3,1,3,2,2,2,2,1,3,1,3,1,2,2,2,1,3,1,1,3,1,1,3,2,2,3,2,1,3,1,1,3,2,1,3,1,2,3,1,3	52	104
V_VCC	10,0,0,0,0,0,10,11,10,0,10,11,10,0,11,10,11,0,0,0,0,10,11,10,11,10,0,0,0,10,11,10,10,11,0,0,1,1,0,10,11,10,10,11,0,10,11,10,0,11,10,11	52	84

Table (1) encoding by using chain code methods

After we get the results of the encoding is clear that the best way is encoded V_VCC note that whenever the shape be more regularly gave better results on the contrary of the other methods that give consistent results for all shapes.

4. The proposed system

The proposed system is lossless image compression, mainly based on two approaches of chain code representation to increase efficiency of algorithm, according to the comparison in Table1; the approach of V_VCC is considered the best in terms of compression ratio, which is used to encode boundaries of the segments, the goal from segmentation is to remove the ***spatial redundancy*** from images.

The second type FCCE is used to encode values that do not form a segment the reason of using FCCE is the ease and flexibility to encode the values in any form if they were connected.

The proposed system consists of two parts, (Compressed and Decompressed), as depicted in figure (7)

The first part consists of four steps:

The first step: finding frequencies of values and put them in a vector then arrange them in descending order, the aim of this step is to increase the efficiency of the algorithm in terms of time and compression ratio, the algorithm starts with value that has highest frequency and ignores the values of the frequencies which does not meet the threshold condition, this step represents the modeling.

The second step: removing spatial redundancy by segmentation process, it begins from the value (VL) that has highest frequency then we take mask $3 * 3$ and VL be the center of mask, looking at the values adjacent for VL to determine values equal to VL and then the mask will be moved to next neighboring which has fewest number of adjacent values equal to VL, this procedure is to determine the boundary of segment, as depicted in figure (5).

	1	2	3	4	5	6
1	8	8	9	7	5	9
2	7	5	14	14	14	9
3	9	14	14	14	14	10
4	10	14	14	14	10	11
5	12	9	14	14	9	11
6	11	8	9	10	12	9

Figure (5) an example to illustrate detection the boundary of segment

The value of location (2, 3) is 14; it has four adjacent values equal to 14, as follows:

1. Location (2,4) has five adjacent values equal to 14
2. Location (3,4) has eight adjacent values equal to 14
3. Location (3,3) has seven adjacent values equal to 14
4. Location (3,2) has four adjacent values equal to 14

So, the algorithm will add the location (2,3) to the vector_{bndr} and moving to the location (3,2) (has fewest number of adjacent values equal to 14) , which is the center of a new mask and perform the same previous procedure and continue until return to the start point.

In this case we have a closed form, now checking the similarity of the internal values after verification from the similarity of internal values, the boundary is encoded based on the values stored in the vector_{bndr} using V_VCC and adding (V_VCC set, start point, color value) to the compressed file then replacing the boundary and internal of the value VL with special value (SV) to become highlighted value.

Then look for another segment of the same color value, and so on, after completing of value VL the algorithm, will select another value which has less frequency and perform same previous procedure.

The third step / remained values that do not form a segment is encoded by using the second approach of chain code (FCCE) for more details, see Ref. (Tawfiq and Fanar 2014).

The fourth step / removing special values from original matrix and shifting the remained values, which leads to reduce the size of matrix, the aim of this step is to generate new contexts to the

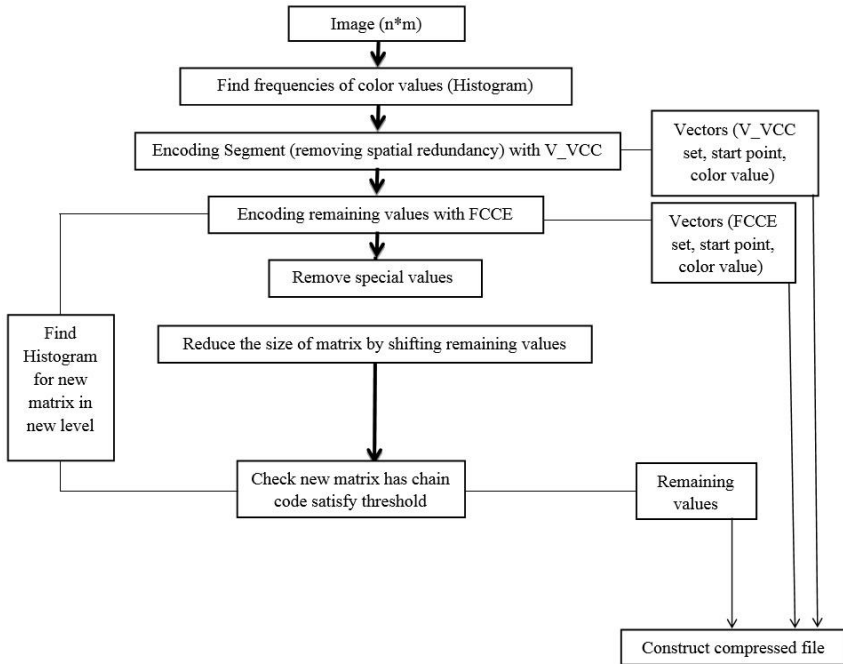


Figure (7) Block diagram (encoding part)

5. Algorithm and experimental results

5.1 Compressed Algorithm:

- Input : Image file (8 bpp – 256 colors)
- Output : Text file

Begin

Level=1

Find frequencies for color values (Histogram)

Sort the vector of histogram (descending order)

Index = 0

Repeat

VL= histogram (index)

Determine boundary of segment for VL

Check boundary of segment

If checkb=true then

Begin

Check internal values of segment

```
If checki=true then
  Begin
    Encoding boundary of segment with V_VCC
    Replace (boundary and internal values) with special
value (-1)
    Add (V_VCC set, start point and VL) to compressed
file
  End;
End;
Index = index +1
Until (index = n)
Repeat
Find histogram
Index=0
  Repeat
    VL=histogram (index)
    Select start point for VL
    Obtain (FCCE) for VL
    If (FCCE (VL)) satisfy threshold then
      Begin
        Add (FCCE set, start point and VL) to compressed file
        Replace location of VL witch correspond to FCCE set with
special value (-1)
      End;
      Index = index +1
    Until (index = n)
  Remove special value from matrix
  Shift remaining values (the size of matrix will be reduce)
  Until matrix has not chain code satisfy threshold condition
End.
```

5.2 Experimental Results

Experiments are performed on 3 color images with 256 colors (8 bpp) as given in Figure (8) to verify the proposed system:

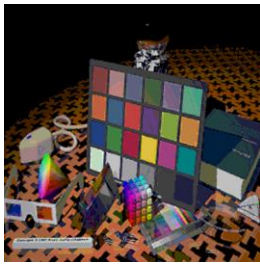


Image 1



Image 2



Image 3

Figure (8) Experimental images

After proposed system is applied on the images above with size of 512*512 we get the following results as depicted in Table (2).

Image	Bytes of image	Bit rate (bpp)
Image 1	262144	3.24
Image 2	262144	3.57
Image 3	262144	4.03

Table (2) compression results of images size 512*5126.

Conclusions

In this research, we have proposed an efficient lossless image compression based on two types of chain code to increase compression ratio based on the experimental results, we note the following points:

- The number of levels have an impact on the required time to compress the image.
- It is possible identify a specific number of levels.
- The segmentation process will be implemented in the first level only that why it's rare to get a segment in other levels because the process of shifting.
- The proposed system can be applied on images (24 bit) by separating the images to 3 matrixes (R, G and B) then compressing each one separately.
- The required time for decompressed images does not exceed one second; it is just process of returning of

values to their original positions in the image depending on V_VCC and FCCE that consider as indicators.

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