An adaptive algorithm for conversion from quadtree to chain codes

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Abstract

An adaptive algorithm is presented for converting the quadtree representation of a binary image to its chain code representation. Our algorithm has the advantage of constructing the chain codes of the resulting quadtree of the Boolean operation of two quadtrees by re-using the original chain codes. This algorithm is adaptive because it can adjust the total number of internal nodes to be stored and retrieve it later in the reconstruction stage. The algorithm possesses parallelism and is suited for pyramid architecture. Our algorithm requires time $O(H + L)$ in sequential and time $O(N)$ in parallel, where $H$ is the height of the quadtree, $L$ is the length of the chain code sequence generated, and $N$ is the size of the input image.

Keywords: Image representation; Chain code; Quadtree; Parallel algorithm; Adaptive conversion

1. Introduction

Representation and manipulation of digital images are two important tasks in image processing, pattern recognition, pictorial database, computer graphics, geographic information systems, and other related applications. Quadtree is one of the compact hierarchical data structures in representing a binary image [1]. It is constructed by successively sub-dividing an image into four equal-size sub-images in the NW (northwest), NE (northeast), SW (southwest), and SE (southeast) quadrants. A homogeneously colored quadrant is represented by a leaf node in the tree. Otherwise, the quadrant is represented by an internal node and further divided into four subquadrants until each subquadrant has the same color. The leaf node with a black (white) color is called the black (white) node and the internal node is called the gray node. An example is shown in Fig. 1 where the alphabet labels are explained in Section 5. There are two widely-used representations of quadtrees. A pointer-based quadtree uses the standard tree representation. A linear quadtree can be either a preorder traversal of the nodes of a quadtree or the sorted sequence of the quadtree's leaves.

The quadtree's data structure has been applied to the representation of maps in the geographic information systems (GIS) successfully [2]. A fundamental operation is to overlay two maps with the union, intersection, or difference. Some research areas are the parallel computational models of the quadtree algorithm [3,4].

Another widely used method to represent digital images is contour representation. A binary image may contain multiple regions. The contour of each region is represented by a starting point and a sequence of moves around the borders. Freeman chain code [5] is one of the common coding techniques. It was developed not aiming at efficient image compression, but rather at making the processing and analysis simpler. The chain code moves along a sequence of border pixels based on 4- or 8-connectivity. The direction of each movement is encoded by using a numbering scheme, such as $\{i \mid i = 0, 1, \ldots, 7\}$ denoting an angle of $45^\circ$ counter-clockwise from the positive x-axis. The chain codes can be viewed as a connected sequence of straight line segments with specified lengths and directions.

Basically, the chain coding algorithm is a sequential approach of contour tracing. It traces the border pixels one-by-one and generates codes by considering neighborhood allocation. A single-pass mid-crack coding
algorithm [6] was developed to use a 3 × 3 template for the code extraction in a raster-scan fashion. It only requires a single row-by-row scan to generate all the code sequences even for a binary image composed of multiple regions. It has an advantage of detecting the spatial relationship between object regions. This algorithm can generate chain codes or mid-crack codes by adopting different look-up tables [7]. Based on the similar concept of extracting the chain codes first and then linking them together, Zingaretti et al. [8] presented another fast chain coding method.

Since both quadtree (region representation) and chain code (contour representation) have their own advantages, it is of interests to develop a method of converting from one representation to another. Existing algorithms from chain code to quadtree conversion were found in Refs. [9,10]. The conversion from quadtree to chain code was reported by Dyer et al. [11] and Kumar et al. [12]. Both algorithms use the same approach as follows.

1. Find the starting node.
2. Extract the chain code of the current node.
3. Search the neighbors of the node in step 2.
4. Mark the edge between two nodes in steps 2 and 3. If the edge has been visited, the process is terminated. Otherwise, go back to step 2.

Our adaptive conversion algorithm from quadtree to chain code is based on the 8-neighbor traversal algorithm [13]. When the traversal algorithm reaches a node at the lowest level, we perform a merging process for chains that come from different nodes in each level. For example, an internal node will have four children and all chains that come from them are merged. The result of the merging process is a chain set. Then we return the resulting chain set to the parent of this internal node. Finally, we will have a chain set which represents the whole image when the root node is reached.

Unlike the existing conversion algorithms, our new algorithm has the superior capability of processing the multiple regions of an input image without additional effort. If we want to extract the chain code information from the set union, intersection, or difference of two quadtrees, our algorithm has the capability of reconstructing the new chain code information easily. This algorithm is suited for parallelization in pyramid architecture because it is based on the recursive function call from one node to its four subquadrants.

In the next section, we will present traversal algorithm for the pointer-based quadtree. Chain coding for the traversal algorithm is described in Section 3. In Section 4, the adaptive conversion algorithm based on the chain code contour reconstruction is discussed. An example is illustrated in Section 5. Analysis of the algorithm is given in Section 6. Finally, conclusions are made in Section 7.

2. Traversal algorithm for pointer-based quadtree

Samet [14] proposed a generic top-down quadtree traversal algorithm in which each node and all 8 neighbors (orthogonally and diagonally adjacent) are visited in preorder. Fuhrmann [13] proposed a simple variation of Samet’s algorithm which can be described in the C language style as follows.

```c
struct NODE {
    char color;
    char NODE *nw, *ne, *se, *sw;
};

traverse (root, w, n, e, s, nwc, nec, sec, swc) {
    if (root->color == GRAY) {
        /* traverse the northwest son of root */
        traverse (root->nw, root->ne, root->s, root->sw);
        nwc = se, n = se, root = se, w = se);
        /* traverse the northeast son of root */
        traverse (root->ne, root->nw, root->e, root->sw);
        n = sw, nec = sw, e = sw, root = sw);
        /* traverse the southwest son of root */
        traverse (root->sw, root->se, root->nw, root->se,
```
The tracking of the traversal algorithm for the 8-neighbors of the SE child of the most northwest quadrant in a 32 image

Table 1
The tracking of the traversal algorithm for the 8-neighbors of the SE child of the most northwest quadrant in a 32 × 32 image

<table>
<thead>
<tr>
<th>Neighbor type</th>
<th>1st stage</th>
<th>2nd stage</th>
<th>3rd stage</th>
<th>4th stage</th>
</tr>
</thead>
<tbody>
<tr>
<td>Central pixel</td>
<td>root → nw</td>
<td>root → nw → nw</td>
<td>root → nw → nw → nw</td>
<td>root → nw → nw → nw → se</td>
</tr>
<tr>
<td>West</td>
<td>Null</td>
<td>Null</td>
<td>Null</td>
<td>root → nw → nw → nw → sw</td>
</tr>
<tr>
<td>North</td>
<td>Null</td>
<td>Null</td>
<td>Null</td>
<td>root → nw → nw → nw → ne</td>
</tr>
<tr>
<td>East</td>
<td>root → nw</td>
<td>root → nw → ne</td>
<td>root → nw → nw → ne</td>
<td>root → nw → nw → nw → sw</td>
</tr>
<tr>
<td>South</td>
<td>root → sw</td>
<td>root → nw → sw</td>
<td>root → nw → nw → sw</td>
<td>root → nw → nw → nw → ne</td>
</tr>
<tr>
<td>NW corner</td>
<td>Null</td>
<td>Null</td>
<td>Null</td>
<td>root → nw → nw → nw → sw</td>
</tr>
<tr>
<td>NE corner</td>
<td>Null</td>
<td>Null</td>
<td>Null</td>
<td>root → nw → nw → nw → sw</td>
</tr>
<tr>
<td>SE corner</td>
<td>root → se</td>
<td>root → nw → se</td>
<td>root → nw → nw → se</td>
<td>root → nw → nw → nw → se</td>
</tr>
<tr>
<td>SW corner</td>
<td>Null</td>
<td>Null</td>
<td>Null</td>
<td>root → nw → nw → nw → sw</td>
</tr>
</tbody>
</table>
nw_child_assign (root, w, n, e, s, nwc, nec, sec, swc, &rootp, &wp, &np, &ep, &sp, &nwcp, &necp, &secp, &swcp);

link_nw = traverse (x1, y1, (x1 + x2)/2, (y1 + y2)/2, ctype, pptype, NW_TYPE, current_level + 1, tree_level, rootp, wp, np, ep, sp, nwcp, necp, secp, swcp);

link_ne = traverse (x1, (x1 + x2)/2, y1, (y1 + y2)/2, ctype, pptype, NE_TYPE, current_level + 1, tree_level, rootp, wp, np, ep, sp, nwcp, necp, secp, swcp);

link_sw = traverse (x1, y1, (x1 + x2)/2, (y1 + y2)/2, ctype, pptype, SW_TYPE, current_level + 1, tree_level, rootp, wp, np, ep, sp, nwcp, necp, secp, swcp);

link_se = traverse (x1, (x1 + x2)/2, y1, (y1 + y2)/2, ctype, pptype, SE_TYPE, current_level + 1, tree_level, rootp, wp, np, ep, sp, nwcp, necp, secp, swcp);

/* merge the chain sets from the four children */
flink = merge (link_nw, link_ne, link_sw, link_se);

else {
    /* perform the chain coding */
    flink = chain_coding (root, w, n, e, s, nwc, nec, sec, swc, x2, y2);
}

return (flink);

/* follow Fuhrmann’s traverse algorithm to assign 8-neighbor */
void nw_child_assign ()
{
    /* input nodes */

    if (root->color ! = GRAY)
        *rootp = *ep = *sp = *secp = root;
    else {
        *rootp = root->nw;
        *ep = root->ne;
        *sp = root->sw;
        *secp = root->se;
    }

    if (w = = NULL)
        *wp = *swcp = DUMMY_WHITE_NODE;
    else {
        if (w->color ! = GRAY)
            *wp = *swcp = w;
        else {
            *wp = w->ne;
            *swcp = w->se;
        }
    }

    if (n = = NULL)
        *np = *necp = DUMMY_WHITE_NODE;
    else {
        if (n->color ! = GRAY)
            *np = *necp = n;
        else {
            *np = n->sw;
            *necp = n->se;
        }
    }

    if (nwc = = NULL)
        *nwcp = DUMMY_WHITE_NODE;
    else {
        if (nwc->color ! = GRAY)
            *nwcp = nwc;
        else {
            *nwcp = nec->se;
        }
    }

Our method uses “virtual” leaves of the tree by “pretending” that the child of a leaf is a leaf of the same color. Even though a node does not have a neighbor, the algorithm provides a virtual neighbor which preserves
the correct color. For example, a leaf node which represents a pixel in the southern boundary of an image has no south neighbor- hood. This traversal algorithm will pretend that it has “white” colored neighbors in s, swc and sec. The assignment for a node’s 8 neighbors is done in the xx-child-assign procedure, where xx denotes nw, ne, sw and se. If the color of a neighbor node is not gray, we use the same node instead for being no descendant of that node. During the traverse, if it is a white node, the further traverse will be skipped for being no contour generation. If it is a black node and its grandparent’s color is black, then this black node may not generate a contour. The criteria are quite simple and described as follows:

1. The black node is NW_TYPE and its parent is SE_TYPE.
2. The black node is NE_TYPE and its parent is SW_TYPE.
3. The black node is SW_TYPE and its parent is NE_TYPE.
4. The black node is SE_TYPE and its parent is NW_TYPE.

Therefore, this algorithm will only traverse the black nodes which lie on the boundary of objects. When the traverse reaches the deepest tree level, the chain_coding procedure is performed. The location (coordinates) of the pixel is computed and passed down from each traversal step. The X coordinate is equal to \( x_2 \) and the Y coordinate is equal to \( y_2 \). The coordinates start from left to right and top to bottom. The coding procedure is performed only on the leaf nodes including all virtual leaf nodes of the quadtree at the tree level (deepest one). For any other node, the merging procedure is performed. The input of the merging procedure is the coding result or the merging result of the subtree of that node.

3. Chain coding for quadtree traversal algorithm

To help readers understand the proposed chain coding algorithm, we first briefly describe our previous work [6,7,15]. Let “Headcode” denote the first code of a chain and let “Thead” denote the code which concatenates with the tail of a chain. Let “Head coordinates” denote the location where the first code points to and let “Tail coordinates” denote the location where the Thead stands on. Let “codelink” be a chain code created for an object pixel. Codelinks will join together in the merging processing and turn to a chain. A chain set is composed of chains.

A set of \( 3 \times 3 \) masks is able to contain all varieties of chain codes in the neighborhood of an object pixel. According to the permutation, a look-up table is set up for the code generation. During the scanning of the input image, if an object pixel is visited, a weighted window is convolved with the \( 3 \times 3 \) neighborhood centered at that pixel. This \( 3 \times 3 \) window, which is incorporated with different weights at each element exploring the presence of eight neighboring locations, is used to calculate the index value of the look-up table. Then, a series of operations are applied to concatenate these individual codelinks with an existing chain or to create a new chain. The conditions for the concatenation are based on the matching of the coordinates and the chain codes of the joining point.

To convert from the quadtree to chain code, the link-list data structures which are convenient in the merging process are used. The nine link-list data structures in Fig. 2 are used to describe a chain set. There are four shared boundaries for a sub-image in a quadrant: north, south, east and west boundaries. Link-lists \( H_n, H_s, H_e \) and \( H_w \) store the Head coordinates and the Headcode of a chain. Link-lists \( T_n, T_s, T_e \), and \( H_w \) store the Tail coordinates and the Thead of a chain. The subscripts \( n, s, e \) and \( w \) denote the north (upper), south (lower), east (right), and west (left) boundaries. The last link-list \( L_c \) is for the code strings and information. The subscript \( c \) denotes the chain.

When a codelink is created, we must find out which shared boundary its Head and Tail are located at. Later, the merging procedure can be performed correctly. When the codelink is 0, it points to the east direction. It will connect to a chain whose Thead is located at its eastern shared boundary. Therefore, its Head coordinates of the
codelink are added to $H_e$. When the codelink is 4, it points to the west direction. It will connect to a chain whose Head is located at its western shared boundary. Therefore, its Head coordinates are added to $H_e$. From the similar concept, if the codelink is 2, we add them to $H_e$. If the codelink is 6, we add them to $H_e$. When the codelink is 1, 3, 5, or 7, it points to the northeast, northwest, southwest and southeast direction, respectively. The shared boundary type where the connection occurs will depend on its pixel location. Before an example illustration, we need to describe the order in the merging procedure to make it easier to understand the assignment of the Head coordinates. As aforementioned, a non-leaf node has four children. When a non-leaf node is merged, we first merge the sub-trees of its $nw$ and $ne$ children. The merging direction is the $x$-direction. Then, it follows by the merging the sub-trees of its $sw$ and $se$ children. The merging direction is the $y$-direction.

If the codelink is 1, it points to the northeast direction. If the pixel which generates the codelink is a $SE$ child in a quadrant, the codelink will connect on the eastern shared boundary. Therefore, its Head coordinates are added to $H_e$. If the pixel which generates the codelink is a $NW$ or $SW$ child in a quadrant, the codelink will connect on the northern shared boundary. Therefore, its Head coordinates are added to $H_e$. If the pixel which generates the codelink is a $NE$ child in a quadrant, the codelink will connect whether on the eastern shared boundary or the northern boundary. The exact shared boundary type where the connection will occur depends on the pairs which are the parent type or the pseudo parent type and the type of the node. The pseudo-parent type is an inheritance based on some special combinations of the node and its parent. For example, if the type of a node is $SW$ and that of its parent is $NE$ but its pseudo-parent type is $NW$, then it will pass down the parent type $SW$ and pseudo-parent type $NW$ to its child. If the situation is not in these rules, the pseudo-parent type is equal to the parent type. These rules are shown in Table 2. From the similar deduction, the shared boundaries where the Tail of codelinks are located are classified. Both of them are shown in Table 3. Codelink “1” is used as an example to illustrate the classification in Fig. 3. In the figure, we only show the $NW$ portion (16 x 16) of a 32 x 32 input image. All other cases can be derived by the same concept.

The order of the codelink creation on a pixel has been shown in Fig. 4. The order granted to each codelink

### Table 2

<table>
<thead>
<tr>
<th>Parent type or pseudo-parent type</th>
<th>Node type</th>
<th>Pseudo-parent type passed down to child</th>
</tr>
</thead>
<tbody>
<tr>
<td>NW</td>
<td>SW</td>
<td>NW</td>
</tr>
<tr>
<td>NE</td>
<td>SE</td>
<td>NE</td>
</tr>
<tr>
<td>SW</td>
<td>NW</td>
<td>SW</td>
</tr>
<tr>
<td>SE</td>
<td>NE</td>
<td>SE</td>
</tr>
</tbody>
</table>

### Table 3

Classification of the shared boundary occurrence for the Head and Tail of a codelink

<table>
<thead>
<tr>
<th>Headcode, Thead</th>
<th>Child type in a quadrant</th>
<th>Parent or pseudo-parent type</th>
<th>Add to shared boundary</th>
</tr>
</thead>
<tbody>
<tr>
<td>0; 0</td>
<td>NW, NE, SW, SE</td>
<td>Don’t care</td>
<td>$H_e; T_w$</td>
</tr>
<tr>
<td>1; 5</td>
<td>NW, SW</td>
<td>Don’t care</td>
<td>$H_e; T_w$</td>
</tr>
<tr>
<td></td>
<td>NE</td>
<td>NW, NE, SW</td>
<td>$H_e; T_w$</td>
</tr>
<tr>
<td></td>
<td>NE</td>
<td>SE</td>
<td>$H_e; T_y$</td>
</tr>
<tr>
<td></td>
<td>SE</td>
<td>Don’t care</td>
<td>$H_e; T_y$</td>
</tr>
<tr>
<td>2; 2</td>
<td>NW, SW, SW, SE</td>
<td>Don’t care</td>
<td>$H_e; T_y$</td>
</tr>
<tr>
<td>3; 7</td>
<td>NE, SE</td>
<td>Don’t care</td>
<td>$H_e; T_y$</td>
</tr>
<tr>
<td></td>
<td>NW</td>
<td>NW, NE, SE</td>
<td>$H_e; T_w$</td>
</tr>
<tr>
<td></td>
<td>NW</td>
<td>SW</td>
<td>$H_e; T_w$</td>
</tr>
<tr>
<td>4; 4</td>
<td>NW, NE, SW, SE</td>
<td>Don’t care</td>
<td>$H_e; T_y$</td>
</tr>
<tr>
<td>5; 1</td>
<td>NE, SE</td>
<td>Don’t care</td>
<td>$H_e; T_y$</td>
</tr>
<tr>
<td></td>
<td>SW</td>
<td>NE, SW, SE</td>
<td>$H_e; T_w$</td>
</tr>
<tr>
<td></td>
<td>SW</td>
<td>NW</td>
<td>$H_e; T_w$</td>
</tr>
<tr>
<td>6; 6</td>
<td>NW, NE, SW, SE</td>
<td>Don’t care</td>
<td>$H_e; T_w$</td>
</tr>
<tr>
<td>7; 3</td>
<td>NW, SW</td>
<td>Don’t care</td>
<td>$H_e; T_w$</td>
</tr>
</tbody>
</table>
generated on a pixel will make Head and Tail in order in the link list as many as it could be. From our observation, some particular cases do not produce the resulting H_s, H_w, T_s or T_w in order. An extra swapping procedure is needed for the adjustment. The swapping procedure is to swap the two related Heads or Tails after the codelinks are created. It restores the sequence to make it in order. H_s or H_w will not be in order when a pixel generates two codelinks and both Heads are on the same link list H_s or H_w. The reason is that we grant the creation of codelinks 5 and 7 with the higher priority than 1 and 3. When codelinks 5 and 3 are assigned to H_w, or 7 and 1 are assigned to H_s for a pixel, the sequence of the Head coordinates will not be in order in the vertical direction. Therefore, the swapping procedure is necessary. We show all the exception cases in this category in Table 4.

Table 4

<table>
<thead>
<tr>
<th>Headcode</th>
<th>Child type in a quadrant</th>
<th>Parent or pseudo-parent type</th>
</tr>
</thead>
<tbody>
<tr>
<td>3 &amp; 5</td>
<td>NW</td>
<td>SW</td>
</tr>
<tr>
<td></td>
<td>SW</td>
<td>NW</td>
</tr>
<tr>
<td>1 &amp; 7</td>
<td>NE</td>
<td>SE</td>
</tr>
<tr>
<td></td>
<td>SE</td>
<td>NE</td>
</tr>
</tbody>
</table>

Fig. 5. The patterns that need the extra swapping procedure.

Table 4. The other category is for T_s and T_w. The patterns which need the extra swapping on T_s or T_w are shown in Fig. 5.

By using this approach, the Head and Tail coordinates of the open chains are in sequence in the link-list structure. When the quadtree traversal algorithm in Section 2 reaches the leaf nodes, all 8 neighbors are known. From their color we can calculate the index value of the look-up table for that pixel (leaf node). Then the chain codes are extracted and placed into a chain set. When the chain coding procedures for all the children of the same parent are finished, we merge the chain sets together. Basically, when two subtrees, named A and B, are merged together, we will merge the Head of subtree A to that of B. Then it follows by the Tail of subtree A to that of B. If a chain is closed after merging, it is added into the chain set C of the link-list L_c. Otherwise, the open chain is added into the chain set O of the link-list L_c. The same process is performed recursively from the leaf nodes back to the root node.

4. Adaptive level-based chain coding retrieval

As aforementioned, the Boolean operation of two quadtrees is the fundamental operation for GIS. If we need the chain code information of the resulting quadtree, we must perform the chain coding procedure for the new constructed quadtree. If we compare the new quadtree with the old one, it may not have great difference. This task can be simplified by constructing the chain code information of the new quadtree from that of the old one.
The link-list data structure used to describe the subtree of a node is simple and easy to maintain. During the merging procedure, we can save the intermediate result of all the nodes which are smaller than a given level. This information will be retrieved when we want to find the chain code contour of the new constructed quadtree. This step will avoid lots of unnecessary re-coding processes. During the Boolean operation of two quadtrees, if the color of a node in the original quadtree is changed, we will mark it and all its ancestors are changed. After the Boolean operation is finished, we perform the chain coding algorithm. If a node is not marked, we need not go further. Instead, we retrieve the link-list information from what we have saved before. If a node is marked, we track its four children to see whether any of them is marked. This process is performed recursively until the leaf node is reached or all the four children of the node are not marked.

Since it is impossible to save all the nodes' information, we can save a portion of nodes instead. By given a level number, we can save all the information of nodes which are less than or equal to that level number. During the chain coding of the new quadtree, if a node is not marked and its level number is less than or equal to the given level, we can retrieve their chain codes directly. If its level number is larger than the given level, we must track all its descendants. This approach is adaptive. It the hardware has plenty of memory, we should use a small level number. Otherwise we use a big number.

5. Experimental results

We use Fig. 1 as an example to illustrate to apply our conversion algorithm. Recall that the chain coding process is performed on the leaf nodes or the virtue leaf nodes in a quadtree. The result of the merging process for each internal node and leaf with black color is shown in Table 5. C, E, F, J and K are leaves with black color but only J and K are individual pixels in the input image. B, D, G, H and I are internal nodes. Root node is marked as A which has two closed chains. They are represented as (0, 1) 0246 and (3,4) 44466600001044432102465, where (x, y) are column and row coordinates of the starting pixel in the chain code description and the numbers follow denote the code sequence of the contour. The code sequence is read from left to right.

6. Analysis

Let the size of an image be $N \times N$ and the number of boundary pixels be $O(B)$. The height of the quadtree (including virtual leaves) is $H$, where $N = 2^h$. Each chain coding and merging operation costs time $O(1)$. The time requiring merging of the chains of two quadrants is dependent on the number of joints in these two quadrants. There is no need to sort or search for the two pairs of the joint (the boundary pixel) since the Heads and Tails are in order. In the sequential mode, the total number of joints in the image is in the same order of boundary pixels and that is the time needed in the traverse of nodes. The complexity of our algorithm is dependent on the height of the traverse plus the operation of the boundary pixels. Therefore, it is $O(H + B)$.

For a pyramid architecture with $O(N)$ processors and each processor handling a subquadrant from its parent, the traverse of the tree needs time $O(H)$ in parallel. The chain coding operation is $O(1)$. The merging operation in a processor of level $l$ requires time which is in the same order of the number of open chains. The worst case occurs when the order open chains' number is equal to the order of the side's length of the quadrant in level $l$. If

<table>
<thead>
<tr>
<th>Node</th>
<th>Head</th>
<th>Tail</th>
<th>Starting pixel</th>
<th>Chain code sequence</th>
</tr>
</thead>
<tbody>
<tr>
<td>C</td>
<td></td>
<td></td>
<td>(0, 1)</td>
<td>0246</td>
</tr>
<tr>
<td>B</td>
<td></td>
<td></td>
<td>(0, 1)</td>
<td>0246</td>
</tr>
<tr>
<td>E</td>
<td>$H_e$ (3, 4)</td>
<td>$T_e$ (4, 3)</td>
<td></td>
<td>02465</td>
</tr>
<tr>
<td>D</td>
<td>$H_e$ (3, 4)</td>
<td>$T_e$ (4, 3)</td>
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the number of the pixels of a quadrant in level \( l \) is \( n \times n \), the time up for the merging operation will require \( O(n) \). If we sum up the merging time for all levels in parallel, which is \( 2^H + 2^{H-1} + \cdots + 2^2 \), it will be \( O(2^H) \) or \( O(N) \).

7. Conclusion

In this paper, an adaptive algorithm for converting the quadtree representation to the chain code representation is presented. Different from other algorithms which need to search for the neighbors of the current node, the algorithm is based on traversing the quadtree. Therefore, it is recursive and has the parallelism. It can be easily implemented in a pyramid architecture. Besides, we introduce an adaptive method to construct the chain codes for the resulting quadtree of the Boolean operation of two quadtrees by re-using the original chain codes. This adaptive method is quite useful and significantly speeds up the conversion in all kinds of applications.

References


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