Fast Detection of Copy Move Image using Four Step Search Algorithm

Yong-Dal Shin*, Yong-Suk Cho**

ABSTRACT

We proposed a fast detection of copy-move image forgery using four step search algorithm in the spatial domain. In the four-step search algorithm, the search area is 21 (<10 ~ +10), and the number of pixels to be scanned is 33. Our algorithm reduced computational complexity more than conventional copy move image forgery methods. The proposed method reduced 92.34% of computational complexity compared to exhaustive search algorithm.

Key words: Copy-Moved Forgery Image Detection, Four Step Search Algorithm

1. INTRODUCTION

Recently, Photo editing softwares such as digital cameras, Paintshop Pro, and Photoshop digital can create counterfeit images easily[1-12]. Various techniques for detection of tamper images or forgery images have been proposed in the literature [1-12]. A form of digital forgery is copy-move image forgery.

Copy-move is one of the forgeries and is used wherever you need to cover a part of the image to add or remove information. Copy-move image forgery refers to copying a specific area of an image itself and pasting it into another area of the same image [1-12]. You can paste part of an image into another part of the image to hide important information or images from the image[9]. To perform this type of paste operation on images that can not be detected by the human eye, copy-forgery is used for the true nature of the image[9]. This technology introduces a correlation between the original image area and the pasted content[9]. To create convincing forgery, you must resize and rotate the paste portion of the image[9]. Fig. 1 shows an example of copy-move forgery. Fig. 1(a) is original image of the Airplane test image. The Fig. 1(b) is copy-move forgery image [1-12].


The most methods of copy-move forgery image detection are divided into overlapping blocks in the search area [1-12]. In order to detect copy-move forgery image of NxN, the copy-move image is divided into (N-B-1)^2 overlapping block size B [6,9-12].

The exhaustive search method of the copy-move forgery image detection needs huge computational complexity [1-4,9-12], not only spatial domain but also frequency domain [11,17,9-12]. The [9-12] proposed a fast detection method of the copy-move forgery image using three step search in the spatial domain [10]. When compared to the frequency domain algo-
Figure 1. Copy-move forgery Airplane image, (a) Original image, (b) copy-move forgery image.

Figure 2. Copy move forgery image by image shifting x and y.

2. THE PROPOSED METHOD

In copy move forgery image, you must compare all the area pairs because they have different shapes and sizes to detect duplicate areas of an image[9]. Most overlapping block methods divide the image into overlapping blocks of fixed size. Blocks are slid by one pixel from the upper left corner of the image to the lower right corner[9]. For MN pixel image, sliding creates (M-B+1)×(N-B+1) blocks[9]. Therefore, detection of copy move forgery image needs huge complexity[10-12].

We have an image \( I(i,j) \) by shifting motion vector \((x,y)\), we can get the copy-move forgery image \( I(P) \), such that[9]

\[
I(P) = I(i-x,j-y)
\]

(1)

Fig. 2 shows copy move forgery image by image shifting motion vector \(x\) and motion vector \(y\). C and P of the Fig. 2 are copy and pasted block respectively.

In order to find a copy-move image region by pixel-matching, it is tested whether each pair of sub-blocks is similar[9].

We proposed a fast detection method of the copy move forgery image using four step search algorithm to reduce computational complexity more than conventional methods in the spatial domain. The Koga et al[17] proposed three step search algorithm in the spatial domain[9-12] It became very popular in the motion estimation for moving pictures because of its simplicity and near optimal performance. The algorithm may be described as[17].

In order to detect copy move forgery image, the four step search algorithm is as follow.

Step 1: Step 1 includes a search based on a 4-pixel / 4-line resolution in a 9×9 search window with nine positions, i.e., a center point corresponding to a zero motion vector such as the dark circle in Fig. 3.

Step 2: Step 2 includes a search based on a 3-pixel / 3-line resolution in a 7×7 search window with nine positions, i.e., a center point corresponding to a zero motion vector such as the white circle in Fig. 3.

Step 3: Step 3 includes a search based on a 2-pixel / 2-line resolution in a 5×5 search window with nine positions, i.e., a center point corresponding to a zero motion vector such as the dark triangular in Fig. 3.
Step 4 : Step 4 includes a search based on a 1-pixel / 1-line resolution in a 3×3 search window with nine positions, i.e., a center point corresponding to a zero motion vector such as the dark rectangular in Fig. 3.

For a maximum displacement window of 10 i.e. \( w=10 \), the number of checking points required is \( [9 \text{ (Step 1)} + 8 \text{ (Step 2)} + 8 \text{ (Step 3)} + 8 \text{ (Step 4)}] = 33 \) check points for search window 21 x21 pixel block.

The computational complexity of four step search algorithm needs 33 checking points.

We used block difference measure (BDM) for matching criterion of copy–moved forgery image detection [10-12]. The BDM expressed equation (2).

\[
BDM = \sum |f(i+i,j+j) - f(i+i+x,j+j+y)| \quad (2)
\]

Where, \( f(i+i,j+j) \) is gray level value at the position \( (i, j) \) of reference 8×8 pixel block image. The \( i=0,1,2,\ldots,N-B-1 \), \( j=0,1,2,\ldots,N-B-1 \), \( ii=0,1,2,\ldots,7 \), \( jj=0,1,2,\ldots,7 \), is gray level value of 8x8 pixel block at the position \( (i+i+x,j+j+y) \) of matching search region image. The \( x \) and \( y \) are motion vectors, which obtained by Four step search algorithm.

The maximum displacement window is 10 for FSS. \( N \) is test image size, we used \( N \) is 256. \( B \) is a pixel block size 8. The \( m \) and \( n \) are divided non-overlapping by 21×21 pixel block in the matching search region image.

BDM is the sum of the differences in gray level values of the reference image block and the matching search image block [10-12].

We used matching criterion (MC) of copy–moved forgery image detection. The MC expressed eq. (3)[10-12].

\[
\text{If } (\text{BDM} = 0) \quad \text{Copy–moved forgery block} \quad (3)
\]

\[
\text{Else} \quad \text{Not copy–moved forgery block}
\]

From equation (3), if BDM was 0, the block is a copy–moved forgery block. This means that the copied block image is the same as the moved block image If BDM was not 0, the block is not copy–move forgery block. This means that the copied block image is different from the moved block image. The flow chart of proposed method shows Fig. 4. From the FIG.3, we find copy move forgery image by MC, BDM using four step search algorithm.

3. EXPERIMENTAL RESULTS

We proposed a new fast detection method for
copy moved image forgery using four step search algorithm in the spatial domain. The proposed method reduces computational complexity compared to various existing methods [10-12].

We used mcar, airplane, house images for computer simulations, the test images 8 bits and 256×256 pixels [9-12,14]. We tampered test images by copy-move one image block over another, in the same image. The matching search region of test images are divide non-overlapping by 21×21 pixel block from first image location (0,0) to last image location (255,255). We used equation (2) to find BDM of copy-moved forgery image based on FSS algorithm. The MC of copy-moved image forgery detection used equation (3). From equation (3), if BDM was 0, the block is a copy-moved forgery block. This means that the copied block image is the same as the moved block image. If BDM was not 0, the block is not copy-move forgery block. This means that the copied block image is different from the moved block image.

In this paper, the computation complexity of FSS algorithm need 33 checking points for 21×21 pixel block. BDM of FSS algorithm is sufficient by 33 checking points instead of 441 checking points (exhaustive search) for 21×21 pixel block to reduce computational complexity.

The Table 1 shows computational complexity results of the proposed method and other methods for test 3 images base on reference [1].

<table>
<thead>
<tr>
<th>Method</th>
<th>Image representation</th>
<th>Block size</th>
<th>Computation complexity</th>
<th>Feature dimension</th>
<th>% Computation complexity (reference [1])</th>
</tr>
</thead>
<tbody>
<tr>
<td>Exhaustive search</td>
<td>Gray level</td>
<td>8 × 8</td>
<td>(256-8-1)²</td>
<td>64</td>
<td>100.0%</td>
</tr>
<tr>
<td>Fridrich[1]</td>
<td>DCT</td>
<td>8 × 8</td>
<td>(256-8-1)²</td>
<td>64</td>
<td>100.0%</td>
</tr>
<tr>
<td>Popescu[2]</td>
<td>PCA</td>
<td>8 × 8</td>
<td>(256-8-1)²</td>
<td>32</td>
<td>50.00%</td>
</tr>
<tr>
<td>Kahn[7]</td>
<td>DWT</td>
<td>8 × 8</td>
<td>(128-8-1)²</td>
<td>64</td>
<td>26.01%</td>
</tr>
<tr>
<td>Shin[11]</td>
<td>Gray level</td>
<td>8 × 8</td>
<td>((256-8-1)/15)² x 25</td>
<td>64</td>
<td>11.65%</td>
</tr>
<tr>
<td>Proposed method</td>
<td>Gray level</td>
<td>8 × 8</td>
<td>((256-8-1)/21)² x 33</td>
<td>64</td>
<td>7.66%</td>
</tr>
</tbody>
</table>

The Table 2 shows three step search to proposed method of computational complexity.

<table>
<thead>
<tr>
<th>Methods</th>
<th>Search block size</th>
<th>Computation complexity</th>
<th>% Computation complexity</th>
</tr>
</thead>
<tbody>
<tr>
<td>Shin[11]</td>
<td>15 × 15</td>
<td>(256-8-1/15)² x 25</td>
<td>100.0 %</td>
</tr>
<tr>
<td>Proposed</td>
<td>21 × 21</td>
<td>((256-8-1/21)² x 33</td>
<td>65.75 %</td>
</tr>
</tbody>
</table>
Table 3. The forgery image detection rate of the proposed method

<table>
<thead>
<tr>
<th>Images</th>
<th>DR</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mcar3</td>
<td>99.89 %</td>
</tr>
<tr>
<td>Airplane</td>
<td>99.63 %</td>
</tr>
<tr>
<td>House</td>
<td>99.22 %</td>
</tr>
<tr>
<td><strong>Average</strong></td>
<td>99.58 %</td>
</tr>
</tbody>
</table>

rate is 99.58% for 3 test images.

Detection rate of copy move forgery image (DR) is expressed by equation (4).

\[
DR = \frac{DPN}{TPN}
\]

(4)

where, detected pixel number of copy move image (DPN), total pixel number of copy move image (TPN).

The Fig. 5, 6, and 7 showed results of proposed method. We showed original images, copy-moved image forgery, detection of copy-moved image forgery in the figure 5, 6, 7. The Fig. 5(a), 6 (a), 7(a) showed original images. The Fig. 5(b), 6 (b), 7(b) showed copy-move images. The Fig. 5(c), 6 (c), 7(c) showed detection image of copy move forgery image. From the Fig. 5(c), 6 (c), and 7(c), left black box is copied, right black box is moved to image forgery block. From Fig. 5, 6, and 7, our algorithm detected copy-moved image forgery by 99.89%, 99.63%, and 99.22%.

4. CONCLUSIONS

In this paper, we proposed a fast detection of copy-move image using four step search algorithm in the spatial domain. In the four-step search algorithm, the search area needs 21 (-10 ~ -10). Our algorithm reduced computational complexity more than conventional copy move image forgery methods. The proposed method reduced 92.34% of computational complexity than exhaustive search algorithm.

REFERENCES


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