

Exposing Digital Image Forgeries Using Feature Extraction and Adaptive Over Segmentation

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ABSTRACT: A novel copy-move forgery detection method using adaptive over segmentation and feature point matching is proposed in this work. The proposed method integrates both block-based and key point-based forgery detection methods. First, the adaptive over segmentation algorithm divides the host image into non-overlapping and irregular blocks adaptively. Then the feature points are extracted using SIFT and matched with each other to locate the labeled feature points which can almost show the suspected forgery regions. To find the forgery regions more precisely, the proposed method provides the forgery region extraction algorithm. This algorithm crops the input image using the located feature points to generate the detected forgery regions. This project is mainly concentrated on improving the detection and localization of duplicated regions using more powerful keypoint-based features. The SIFT algorithm is a powerful feature extraction technique, which extracts features invariant to scale, rotation, and brightness. However, SIFT descriptors are not invariant to mirror reflection. In this context, this work also adopted a more powerful set of keypoint-based features, called MIFT, which shares the properties of SIFT features but also are invariant to mirror reflection transformations. Experimental results show the good performance of the proposed copy-move forgery detection.

KEYWORDS: Copy-move forgery detection, adaptive Over-segmentation, SIFT, MIFT, Matching

I. INTRODUCTION

Nowadays, digital image forgery has been becoming gradually easy to perform. Of the existing kinds of image forgery, one of the common manipulations with digital image is copy-move forgery, which is to paste one or several copied region(s) of an image into another part(s) of the same image. In the past years, many forgery detection methods have been proposed for copy-move forgery detection. During the copy and move operations, some image processing methods such as rotation, scaling, blurring, compression, and noise addition are sometimes applied to make convincing forgeries. Because the copy and move parts are copied from the same image, the noise component, color character and other important properties are compatible with the rest of the image; some of the forgery detection methods that are based on the related image properties are not applicable in this case.

The most apparently obvious way of detecting copied and pasted regions in the same image would be to authenticate small clusters or blocks of pixels for matches all across the image. However, there are two main issues with this approach. Firstly, this would be a computationally expensive approach, as matching blocks (or other shapes) of pixels would become infeasible with increasing size of the image. Secondly, such an approach would fail in case of minor changes such as addition of noise or multiple image compression. In order to avoid these drawbacks of this direct approach, researchers have established various techniques which can be classified into two main categories: block-based and feature-based.

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The current block-based forgery detection methods divide the host images into overlapping and fixed image blocks; then, the interfered region can be obtained by matching blocks of image pixels or transform coefficients. As an alternative to the block-based methods, feature key point based forgery detection methods were suggested, where image key points are extracted and matched over the whole image to resist some image transformations while recognizing duplicated regions.

II. RELATED WORK

Fridrich et al. [1] suggested the forgery detection method where the input image was segmented into over-lapped rectangular blocks, and from which the quantized Discrete Cosine Transform (DCT) coefficients of blocks were matched to find the tampered regions. Luo et al. [2] was used the RGB color components and direction information as block feature. Li et al. [3] was used Discrete Wavelet Transform (DWT) and Singular Values Decomposition (SVD) to extract image features. Mahdian and Saic [4] calculated the 24 Blur-invariant moments as feature. Bayram et al. [5] used the Fourier-Mellin Transform (FMT) to obtain feature. Wang et al. [6] suggested the mean intensities of circles with different radii around the block center to represent block feature. Ryu et al. [7] used Zernike moments as block feature. Bravo-Solorio and Nandi [8] used the information entropy as block feature. In the Scale Invariant Feature Transform (SIFT) [9] was used to the input images to extract feature points, which are then matched to each other.

Although these methods are effective in forgery detection, they have three main drawbacks: 1) the input image is divided into over-lapped regions, which will cause the computation complexity expensive; 2) the methods cannot deal with significant geometrical transformation of the forgery areas; 3) the host image is divided into regular blocks, which will cause low recall rate. And the existing keypoint based forgery detection methods can somewhat reduce the computation complexity and can be strong against some attacks, the recall results were still poor. To address the shortcomings of the existing methods, we propose a novel copy-move forgery detection scheme using adaptive over-segmentation and feature point matching in this paper. The Adaptive Over-Segmentation algorithm is proposed to adaptively divide the host image into non-overlapping and irregular blocks. Then the feature points are extracted from each block and matched with each other to find the labeled feature points which can approximately indicate the suspected forgery regions. And finally the labeled feature points are processed and the morphological operation is applied to generate the detected forgery regions.

III. PROPOSED METHOD

This section describes the proposed image forgery detection using adaptive over-segmentation [10] in details. Fig. 1 shows the framework of the proposed method for image forgery detection. Firstly, the adaptive over-segmentation method is proposed to divide the input image into non-overlapping and irregular regions. Then SIFT [9,10] is applied into each block to extract feature points as block features which are matched with each other to locate the points which can approximately specify the suspected forgery regions. Finally the forgery regions are detected according to the matched feature points.

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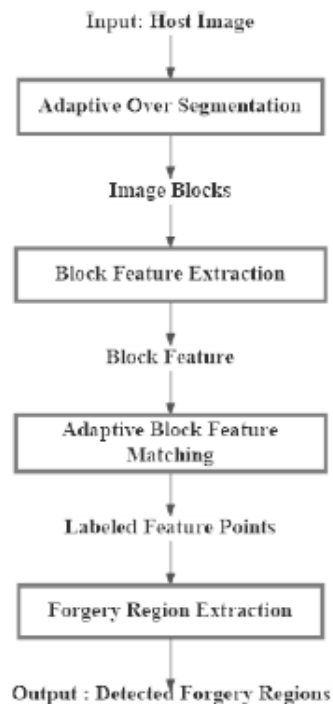


Fig. 1 Framework of the proposed copy-move forgery detection scheme

A. Adaptive Over Segmentation

In order to segment the input image into non-overlapping regions of irregular shape, we used the SLIC algorithm to divide the host image into meaningful superpixels. As a non-overlapping segmentation method, SLIC [10] can reduce the computational expenditures comparing with the overlapping blocking; furthermore, in most of the cases, the irregular and meaningful regions can signify the forgery region better than the regular blocks. However, the initial size of the superpixels in SLIC is hard to determine. When the initial size is too small, it will cause large computation expenses; otherwise, when it is too large, it will cause the forgery detection results not accurate enough. At present, there is no such a good method to determine the initial size in superpixel segmentation algorithms. Therefore, in this paper, we proposed the Adaptive Over-Segmentation method which can define the initial size adaptively based on the texture of the host image and thus can segment the input image into irregular and non-overlapping blocks. In the proposed Adaptive Over-Segmentation method, firstly, the Discrete Wavelet Transform (DWT) is employed into the host image to generate the low frequency and high frequency sub-bands. Then the initial size of the superpixels is calculated with the adaptive block size computation. Finally, with the calculated initial size, the SLIC segmentation algorithm is employed to divide the input image into irregular and non-overlapping image blocks.

B. Block Feature Extraction-SIFT

After the host image is segmented into image blocks, block features are extracted from the image blocks (IB). The traditional block-based forgery detection methods extracted features of the same length as the block features or directly used the pixels of the image block as the block features. However, these features reflect mainly the content of the image blocks, leaving out the location information. Also, these features are not resistant to various image transformations. Therefore, in this project, the feature points are extracted from each image block as block features and the feature points should be robust to various distortions, such as image scaling, rotation, and JPEG compression. The feature points generated using these methods are robust against common image processing operations such as rotation, scale,

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blurring, and compression. Experiments have shown that the results obtained using SIFT are more constant and have better performance compared to other feature extraction methods. Hence, in this proposed scheme SIFT is used for feature point extraction. Therefore, each block feature contains irregular block information and the extracted SIFT feature points.

C. Block Feature Extraction-MIFT

In this study, we improve copy-move forgery detection using keypoint-based features by focusing on the issue of accurate detection and localization of duplicated regions. Exactly, we have made several contributions in this work. The SIFT algorithm is a powerful feature extraction technique, which extracts features invariant to scale, rotation, and brightness. However, SIFT descriptors are not invariant to mirror reflection. To account for this issue, previous approaches proposed extracting SIFT descriptors from horizontally and vertically reflected versions of the original image. In this paper, we have adopted MIFT [11] descriptors that are invariant to mirror reflection transformations. MIFT is one of the auspicious local feature detector based on SIFT, that is robust to mirror reflection also. A Mirror reflected version of an image can be acquired by reversing the axis of the image, hence, in a horizontally reflected image the row order of pixels remains the same, but the column order changes which is shown in Figure 2. As for original descriptors, a fixed encoding policy is accepted, which results in distinct descriptors in different mirror reflection situations for the same feature, as shown in Fig. 2(e) and (f). So in MIFT mirror invariance is achieved by simple descriptor reorganization. This descriptor reorganization first organizes the arrangement of cell order around the interest points. This is done by checking the values of total left pointing (ml) and right pointing (mr) orientations. Based on the winning orientation, column order may change ($ml > mr$) or not. Second, for each cell, it checks whether the order of orientation bins to follow clockwise or anticlockwise direction. Thus MIFT provides a descriptor which is identical in all cases of mirror reflections.

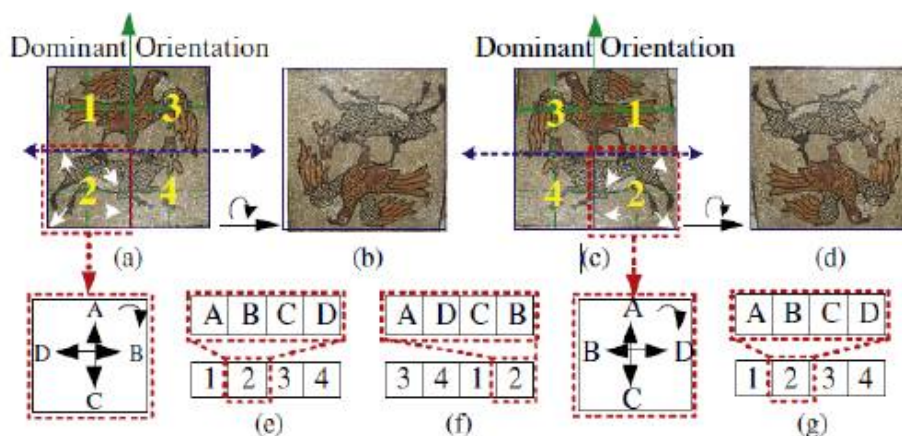


Fig. 2 Different mirror reflections and the concept of MIFT. A simplified SIFT descriptor with 22 cells and 4 oriented bins in each cell is used for illustration. (a) The original image. (b) The combined reflection of (a). (c) The horizontal reflection of (a). (d) The vertical reflection of (a). (e) The original and MIFT descriptor for (a). (f) The descriptor of original version for (c). (g) The MIFT descriptor for (c).

Finally, the algorithm locates the matched blocks through the block features. For that first, the number of matched feature points is calculated, and the correlation coefficient map is generated; then, the corresponding block matching threshold is calculated adaptively; with the result, the matched block pairs are located; and then, the matched feature points in the matched block pairs are extracted and labeled to locate the position of the suspected forgery region. From these labeled feature points, the proposed system finds the matching index from the host image. From that find the lowest point on left and highest point on right is determined and similarly finds the lowest point from top and highest point from bottom. And finally crop the image using these points together the detected forgery region.

IV. EXPERIMENTAL RESULTS

In this section, a sequence of experiments is conducted to evaluate the effectiveness and robustness of the proposed copy-move forgery detection scheme. We have evaluated the performance of the proposed methodology by performing a comprehensive set of experiments with a large database of real images. Comparisons with competitive approaches show that the proposed method can find duplicated regions in copy-move image forgery more accurately, especially when the size of the duplicated regions is small. Below figures shows the result of forgery detection with the input image shown in fig.3.



Fig. 3.(a) Original Image (b) Forged Image

In below figure, fig.4.(b) shows the results after adaptive over segmentation and 4.(c) was the result after feature extraction using SIFT algorithm. Then 4.(d)and (e) was the feature point matching and forgery region extraction results.

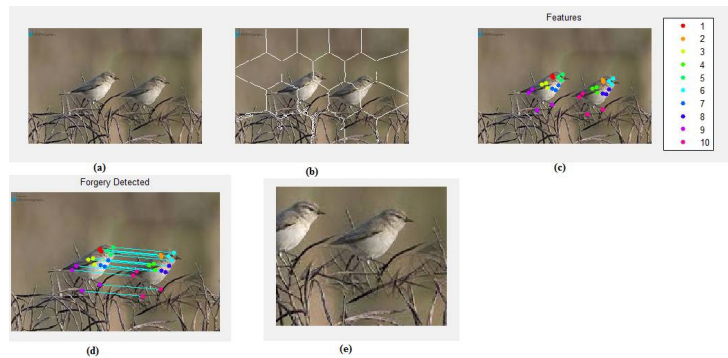


Fig. 4(a) Host Image (b) Segmented Image (c) Feature Extraction using SIFT (d)Feature Matching (e)Forgery Region Extraction

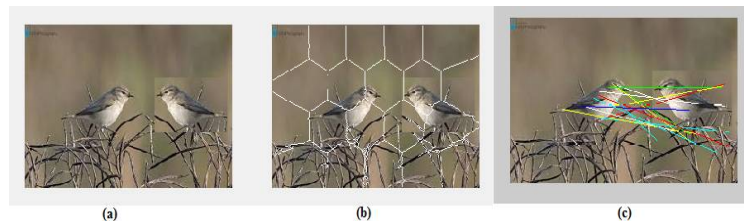


Fig.5 (a) Host Image (b) Segmented Image (c) Feature Extraction using MIFT

Experimental results show that the proposed scheme can achieve much better detection results for copy-move forgery images under various challenging conditions, such as geometric transforms, JPEG compression, and down-sampling. But this method, however, cannot handle reflection. Mirror reflection invariant feature generalizes SIFT by producing mirror reflection invariant descriptors. We have adopted MIFT descriptors in this work to find duplicated regions with or without mirror reflection. Figure 5 shows an example using MIFT in the case of mirror reflection.

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In the above experiments, the two characteristics precision and recall [10] are used to evaluate the performance of the proposed forgery detection scheme. Precision is the probability that the detected regions are relevant, and it is defined as the ratio of the number of correctly detected forged pixels to the number of totally detected forged pixels, as stated in [10]. Recall is the probability that the relevant regions are detected, and it is defined as the ratio of the number of correctly detected forged pixels to the number of forged pixels in the ground-truth forged image. Fig 6 shows the precision and recall rate of the method using MIFT.

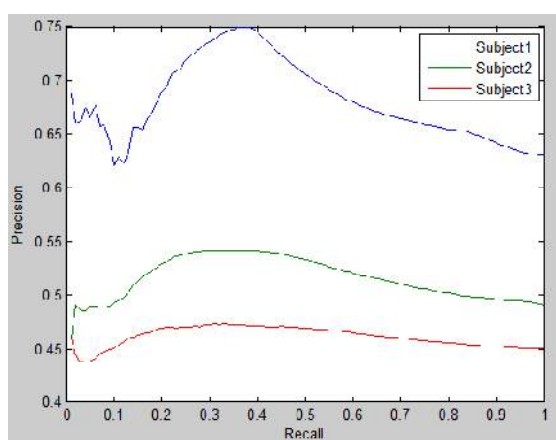


Fig. 6 Precision and Recall Results at the Pixel Level

V. CONCLUSION

In this paper, we have proposed a novel copy-move forgery detection scheme using adaptive over-segmentation and feature extraction. The proposed Adaptive Over-Segmentation algorithm can adaptively divide the input image into non-overlapping and irregular blocks. In each block, the features points are extracted using SIFT and matched to indicate the suspected forgery regions. The proposed methodology employs a new set of keypoint-based features, called MIFT, for finding similar regions in an image. We have performed comprehensive experiments using a large dataset of real images to evaluate the proposed approach. In particular, we have investigated the effect of different transformations in creating the image forgery on detection accuracy. It should be mentioned that like with similar methods employing keypoint-based features for matching, the proposed approach will not work well if the duplicated region corresponds to a flat surface where no interest points can be detected. Future work could focus on applying the proposed forgery detection scheme based on adaptive over-segmentation and feature-point matching on other types of forgery, such as splicing or other types of media, for example, video and audio.

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