

# *Copy-move Image Forgery Detection Based on Gabor Descriptors and K-Means Clustering*

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**Abstract**—At present, popularity of using image as the fundamental media of information is growing. Rapid development of technology brings effective image processing tools available and makes image forgery very easy. As an outcome, it turns into a complicated issue in late time. In that case, validating the legitimacy and integrity of digital images is ending up progressively vital issue. The most challenging region-duplication forgery is made by copying some portion of an image and pasting on different region of the same image. This study proposes an efficient region-duplication forgery detection technique. This research is categorized into segment-based region duplication forgery detection method. The design of the algorithm based on image segmentation and using Gabor descriptors and K-Means clustering. Initially, the image is segmented using normalized cut (NCut) segmentation technique. Then, applied Gabor Filters to extract image features and cluster similar features using K-Means clustering algorithm. Finally, comparing the clustering regions with the given threshold value will decide image authenticity. Experiment results proves the strength of the proposed method against various post-processing attacks such as- rotation, scale, blurring and JPEG compression. A comparison with existing image forgery detection algorithms demonstrates that the proposed algorithm gives better performance.

**Keywords**—image forensics, image forgery detection, passive authentication, region-duplication forgery, copy-move forgery

## I. INTRODUCTION

There is a significant role of digital images in our everyday life. However, image manipulation has become very easy by using powerful software. Without any doubt image authenticity now is a big matter of concern. Two main types of image forensic techniques are available to verify the integrity and authenticity of manipulated image. One is active forensic method [1] and another is passive forensic method [2][3] and they have been explained in the

literature precisely. Watermarking and steganography are two techniques which are used in active methods to insert authentic information into the image. When question arise about the authenticity of an image, then prior embedded authentication information is recalled to prove the authenticity of that image. However, embedding authentication information to an image is very confidential. Only an authorized individual allows to do it or at the time of creating the image, authentic information could be embedded as well. Requirement of special cameras and multiple steps processing of the digital image are two main limitations which made this technique less popular. To avoid these limitations, the researchers are more interested in developing passive forensic technique as these methods utilize image forgery without requiring detailed previous information. The most popular method to construct forged image is copy-move forgery. It refers to copy one part from image, and paste it inside the same image. Sometime before pasting the copied regions, various post-processing operations like scale, rotation, blurring, intensifying or JPEG compression may be applied. The similarity regions between image clusters is used in this proposed method [4]. Firstly, the image is segmented using color regions by Normalized cut (NCut) segmentation technique. Secondly, centroids are identified from each region. Thirdly, image features for each region from the image are extracted by using Gabor descriptors [5]. In the next step, K-Means clustering is applied in order to cluster similar features from all regions. These clusters of features be compared to each other to identify which region has the most similar feature. Here, Euclidean distance calculation be used between each cluster for checking the threshold distance. If the similar feature cluster distance lies between the experimented threshold value, it is defined as tampered region otherwise authentic image. Finally, duplicated region will be localized. The paper is prepared as follows: Section 2 affords a discussion on the existing literature, while Section 3 details the proposed copy-move forgery algorithm.

Section 4 describes the experimented result. Lastly, the paper concludes with section 5.

## II. LITERATURE REVIEW

Various number of region-duplication authentication techniques have been implemented to handle the issues identified with image authentication. All copy-move forgery detection techniques are developed with either block-based [9-13] or keypoint-based methods [14-17].

### A. Block-based Image Forgery Detection

In block-based method, input image size of  $M \times N$  is segmented into overlapping blocks size of  $z \times z$  resulting into overlapping blocks,  $L = (M-z+1) \times (N-z+1)$ . A few features are extricated from each block. Distinctive features are extracted by applying different feature extraction technique such as DCT (Discrete Cosine Transform) [8], DWT (Discrete Wavelet Transform) [9], DFT (Discrete Fourier Transform) [10], PCA (Principal Component Analysis) [11] [12], SVD (Singular Value Decomposition) [13][14], and ZMs (Zernike Moments) [15]. Finally, a comparison is done based on blocks features similarity and distance. After finding the most matched features blocks, copy-move region is identified and localized. Sheng et al. [8] proposed forgery detection algorithm using block-based method. Proposed algorithm used DCT and circle blocking technique for extracting features from the image. Finally, the image which contains singularities within lines is presented by computing ridgelet transformation. Robustness against JPEG compression is the most significant feature of this method. Cao et al. [16] followed block-based method to detect tampered region where DCT feature extraction technique is applied. Initially, DCT is applied onto the divided sub blocks to extract key features by producing quantized coefficients. Threshold value was set prior to match features between closest similar image blocks. This method shows less computational complexity compared to existing methods [24-17] because of reducing dimension of feature vector. Later, similar method and feature extraction technique used by Huang et al. [18]. The big difference in the result with Cao et al.'s DCT-based method [16], because of reducing the false matches rate. Due to low false matches rate, this method becomes powerful against noise and blurring. However, it is not robust for rotation attack and cannot detect multiple forgery. M. Bashar et al., [19] developed more efficient forgery detection method based on DWT and kernel PCA (KPCA) features. Natural images have been used as a dataset in this method. As a consequence of quantitative analysis considering noiseless and uncompressed factor, it is found that the DWT performs well than KPCA in terms of features. On the other hand, in noise and JPEG compression domain, KPCA-based features perform better than DWT. The method shows robustness against noise and JPEG compression attack. Though the method takes too much time and not robust against scaling. It cannot detect multiple forgery.

### B. Keypoint-based Image Forgery Detection Method

Different from block-based methods, features are extracted in keypoints-based method from the image without any type segmentation. Extracted features from every keypoints are compared to find similarities between them. Finally, based on the calculation of matched features, image forgery is detected. SIFT and SURF (Speeded Up Robust Features) are two main keypoints-based feature extraction methods. Somayeh Sadeghi et al., [20] and Diaa M. uliyan et al., [21] worked on keypoints-based technique (e.g. SIFT). Sadeghi et al., proposed SIFT to extract features and searched for similar features based on their Euclidean distance. Both methods are robust against several post-processing attacks; including scale, noise, rotation and JPEG compression. However, inability to detect small forged areas and performance of detection and localization for those forged areas are also questionable. In [21], primary approach of Uliyan et al., was to detect image regions by using Statistical Region Merging (SRM) Segmentation algorithm. Then, the experiment proceeded with applying Angular Radial Partitioning (ARP) and Harris Corner detection method on the image region. Finally, forged regions were detected based on matched keypoints. The method showed less robustness against forged regions with blurring and illumination attacks. Moreover, it shows different result for same image with different resolution. The major drawbacks of the previously mentioned conventional techniques are either not powerful against all post-processing attacks or have high computation time. Therefore, keeping up the low computational time is the most important robustness challenge. To tackle this issue, a new copy-move forgery detection method is proposed where region wise image segmentation is done. Gabor filters are used to extract image features. Afterwards, K-Means clustering, and Euclidean distance calculation facilitated to detect forged region from the suspicious image. Reducing the false matching rate is the most significant task to exhibit the proposed method as more viable compared to conventional methods.

## III. PROPOSED METHOD

In this section, proposed blind copy-move image forgery detection algorithm based on Gabor descriptors and K-Means clustering is explained. Image features can be extracted by using many techniques such as edge detection, and corner detection [22] etc. Though, Gabor filter is chosen as feature extractor in the proposed method and image textures are extracted from each image region. Obviously, texture is the most preferred feature can be get after manipulating image regions every time. Compared to other feature extraction techniques [20][23][24], Gabor filter shows more robustness against scale, rotation, noise and JPEG Compression. It is faster for feature extraction as well. Several steps are followed in the proposed method: Image segmentation based on Ncut (normalized cut)

segmentation technique and localizing the centroids of every region; extracting features by using Gabor descriptors and cluster similar features region based on K-Mean clustering, and lastly the detection of duplicated region in forged image. Proposed method is given as a block diagram in Fig. 1. Brief explanation of each steps is given in the following subsections.

#### A. Image Segmentation Using Normalized Cut and Centroids Detection

The proposed method employs normalized cut (NCut) segmentation to the input image  $I$  of size  $r \times c$ . Image segmentation is done based on color and spatial threshold value with a minimum area of segmentation and smallest NCut value subject to the choice of image as a weighted graph. A weight matrix is calculated based on equation (1).

$$W_{ij} = e^{-\frac{\|F(i)-F(j)\|_2^2}{\sigma_f^2}} * \begin{cases} e^{-\frac{\|X(i)-X(j)\|_2^2}{\sigma_x^2}} & \text{if } \|X(i)-X(j)\|_2^2 < r \\ 0 & \text{otherwise} \end{cases} \quad (1)$$

Where  $X$  denotes the spatial values and  $F$  is the feature in this case. Color values corresponding to point  $i, j$  and  $\sigma$  (sigma) refers to the specified threshold. A  $N \times N$  diagonal matrix  $D$  is formed as  $d_i = (\sum W_{ij})$  with subject to node  $I$  and a generalized Eigen system  $(D-W)_y = \lambda D_y$  is solved to obtain two discrete eigenvectors such that we can bipartition the graph recursively based on the specified minimum threshold value [1][25][26]. Part of the experiment with tampered images and its NCut segmentations are displayed in Fig. 2. After image segmentation, the centroid of each segment is computed and a  $7 \times 7$  block around the centroids for feature extraction and clustering have been proposed.

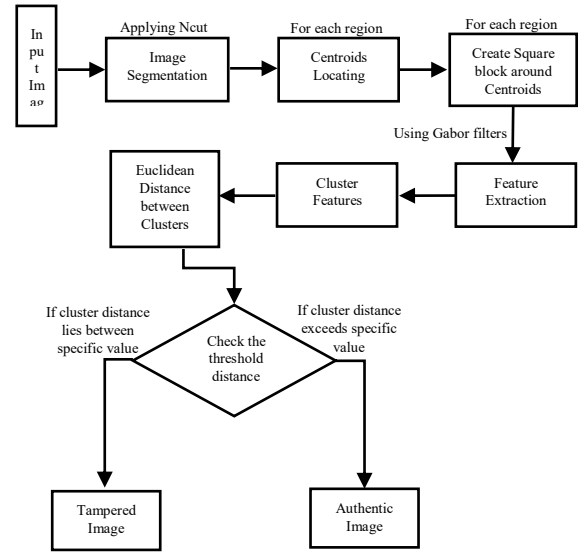


Fig. 1. Block diagram

In view of the proposed research, centroids hold the best significance of an image when compared with other methods [1][18][27][28][29]. The reason is, centroids bring into effective calculation of Euclidean distance between two clusters, also help to find the most similar clusters. Moreover, connection between centroids are utilized in order to calculate their Euclidean distance. For instance, when  $e$  and  $f$  represent two clusters respectively, then the linked centroids can be calculated as:

$$d(e, f) = P\overline{X_e} - \overline{X_f}P_2, \quad (2)$$

where  $\overline{X_e}$  and  $\overline{X_f}$  are defined as:

$$\overline{X_e} = \frac{1}{n_e} \sum_{i=1}^{n_e} X_i, \quad \overline{X_f} = \frac{1}{n_f} \sum_{i=1}^{n_f} X_i, \quad (3)$$

From the above-mentioned equation, the variables  $\overline{X_e}$  and  $\overline{X_f}$  represent the weighted centroids of two different clusters  $i$  and  $j$  [20]. In the proposed method, the block size is experimentally fixed to  $7 \times 7$  around the centroids. Finally, image segmentation using NCut and centroids calculation of each segment is shown as in Fig. 3.

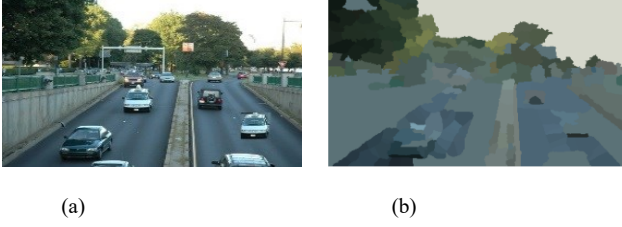


Fig. 2. (a) Copy-move forgery image (b) NCut segmentation



Fig. 3. (a) NCut segmented image (b) Locating centroids

### B. K-Means Clustering of Image Regions Based on Gabor Descriptors

Texture refers as one of the image features which is the normal repetition of an element or pattern on a surface. In Gabor filters, texture features are extracted by analyzing the frequency of the image utilizing frequencies and orientations. In the proposed method Gabor descriptors are applied in each image block to extract texture of each regions [29]. Experiment employed 2-D Gabor filtering technique, as the following:

$$f(x, y, \omega, \theta, \sigma_x, \sigma_y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp \left[ \frac{-1}{2} \left( \left( \frac{x}{\sigma_x} \right)^2 + \left( \frac{y}{\sigma_y} \right)^2 + j\omega(x\cos\theta + y\sin\theta) \right) \right] \quad (4)$$

Where  $x, y$  denotes pixel position,  $\omega$  is the centre frequency to the  $x$  and  $y$  directions.  $\theta$  is for orientation and  $\sigma_x, \sigma_y$  represent standard deviation of the Gaussian along both  $x$  and  $y$  directions.

In the next step of experiment after feature extraction, K-Means clustering algorithm is proposed. It is an unsupervised technique and work as a classifier, which categorizes the input points into various clusters utilizing their in-between principle distance. Random intensities are used to cluster various information around centroids. As the following equation is used:

$$V = \sum_{i=1}^k \sum_{x_j \in S_i} (h(x) - \mu_i)^2 \quad (5)$$

Where  $k$  denotes as clusters. Proposed experiment has clusters value of  $k=2$ ,  $S_i$  (where  $i = 1, 2, \dots, k$ ) and  $\mu_i$  = Centroid of all pixels  $x$  includes the histogram  $h(x) \in S_i$ . At that point, similar clusters and collection of connected

pixels are associated each other to create segmentation map as given below:

$$D = \arg \min_j ||h(x) - \mu_j||^2 \quad (6)$$

The repetition of  $j$  applies over every centroids [28].

### C. Duplicate Region Detection

After examining the feature  $F_i$  of  $S_i$ , the Euclidean distance can be represented as the following:

$$D(f_i, f_j) = \sqrt{\sum_{i=1}^n (f_i - f_j)^2}, i \neq j \quad (7)$$

$D(f_i, f_j)$  represents the most relevant features of matched regions. The forged areas are to be found alongside their matched features [1]. Detection of duplicate region has been illustrated in Fig. 4.

## IV. EXPERIMENTAL RESULTS

The experiment is tested on MICC-F220 dataset [7] which is a popular dataset for the researchers who work on copy-move forgery detection. The dataset contains 220 images where 110 are original and 100 are tampered images. All tampered images in this dataset are post-processed by different attacks like scale, rotation, JPEG compression. The results of this proposed technique are illustrated in Fig. 5, 6, and 7 with sample images, and Table 1. Where Fig. 5 shows that the method can identify region duplication forgery accurately. Fig. 6 shows robustness of the proposed method against combined attack of rotation and JPEG compression. Lastly, Fig. 7 represents the ability of this forgery detection method against scale attack.

The TPR (true positive rate), and FPR (false positive rate) are used to measure the performance of proposed method.

$$TPR = \frac{\text{No. of images detected as forged being forged}}{\text{No. of Forged Images}} \quad (8)$$

$$FPR = \frac{\text{No. of images detected as forged being original}}{\text{No. of Original Images}} \quad (9)$$

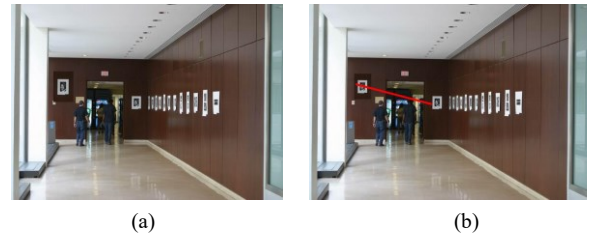


Fig. 4. (a) Forged image (b) Locating forged region



Fig. 5. (a) Sample forged image (b) Detection result

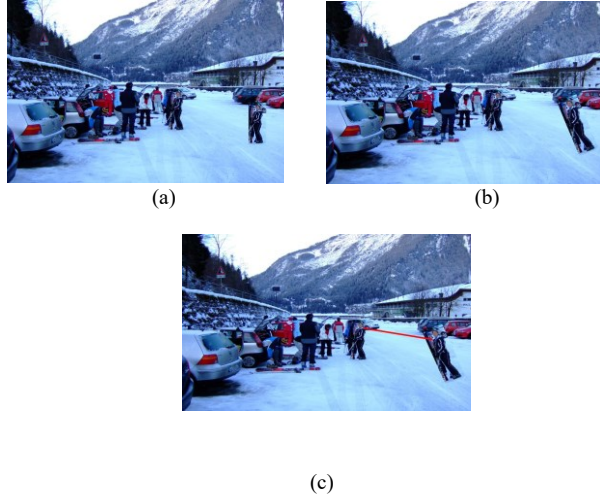
Fig. 6. (a) Forged image (b) Forged image with rotation attack ( $\theta=45^\circ$ ) and recompress (JPEG Q Factor= 60) (c) Locating forged region

Fig. 7. (a) Forged image with scale attack (factor 1.2) (b) Matched centroids between duplicate regions

TABLE I. AVERAGE DETECTION RATE

| Operations       | Quality Factors | 90   | 80   | 70   | 60   | 50   |
|------------------|-----------------|------|------|------|------|------|
| Normal Copy Move | TPR             | 0.93 | 0.93 | 0.92 | 0.90 | 0.90 |
|                  | FPR             | 0.02 | 0.02 | 0.03 | 0.04 | 0.05 |

|          |     |      |      |      |      |      |
|----------|-----|------|------|------|------|------|
| Rotation | TPR | 0.91 | 0.90 | 0.89 | 0.88 | 0.87 |
|          | FPR | 0.09 | 0.10 | 0.11 | 0.11 | 0.21 |

Table 2 shows that the proposed method succeeded with high TPR value (93%) better than Mishra et al [30] and S.Lyu [31]. Moreover, the average FPS of the method is 2.89%, which is lower than Diaa Uliyan, et al., method [17].

TABLE II. DETECTION RESULTS OF PROPOSED METHOD

| Methods  | T <sub>PR</sub> % | F <sub>PR</sub> % |
|--|-------------------|-------------------|
| Segmented-based forgery detection method [17]                                  | 96                | 2.89              |
| Keypoints-based forgery detection based on SURF feature extraction method [30] | 73.6              | 3.64              |
| Forgery detection based on SIFT feature extraction method [31]                 | 89.96             | 1.25              |
| Proposed Method  | 93.18             | 2.81              |

## V. CONCLUSION

This study presented copy-move image forgery detection approach using Gabor descriptors and K-Means clustering. The experiment results demonstrate the ability of the proposed approach against post-processing operations like JPEG compression, scaling and rotation. Moreover, the proposed method gives better results compared with the existing methods in terms of detection rate. TPR is increased to 93.18%, while the FPR is 2.81%. However, the proposed method struggles for detecting multiple forged location in an image and also shows inaccurate result with noise attack. In conclusion, it is to believe that this proposed novel method would be beneficial to some extent in various areas of forensic science.

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