

A New Efficient Forgery Detection Method using Scaling, Binning, Noise Measuring Techniques and Artificial Intelligence (Ai)



Mahesh Enumula, M.Giri, V.K. Sharma

Abstract: In the market new updated editing tools and technologies are available to edit images and with help of these tools images are easily forged. In this research paper we proposed new forgery detection technique with estimation of noise on various scale of input image affect of noise in input image, frequency of images are also changed due to noise, noise signal correlated with original input images and in compressed images quantization level frequency also changed due to noise. We partition input image into $M \times N$ blocks, resized blocks are proceed further, image colors are also taken into consideration, each block noise value is evaluated at local level and global level. For each color channel of input image estimate local and global noise levels are estimated and compared using binning method. Also measured heat map of each block and each color channel of input image and all these values are stored in bins. Finally from all noise values calculate average mean value of noise, with these values decide whether input image is forgery or not, and performance of proposed method is compared with existing methods.

Keywords: Image Processing, Machine Learning, Retouching, Binning, Forgery, Water Marking.

I. INTRODUCTION

With latest development tools and technology, an image can be easily editable, untrained peoples may not in a position to differentiate original and forged images[1], many image processing tools are available to detect such a forgery images, all these tools are giving clues to detect forgery, a window sized boxes will scan entire images based on clues, and all these image forgery detection techniques are suffering with boundary problems. Authors in [3] introduced image locations technique to detect forgery within segments, image localization is done with noise and photo response technique, taken homogeneity of local region to detect forgery based on small clues, and authors[3] combined localization technique and window based frame work to detect forgery images.

Median filtering is widely used in multi-media-based applications like steganography, image forensic, security filtering approach is mostly nonlinear technique and as well as data also preserving using this technique. Earlier researcher on image processing proposed various types of techniques to handle images based on median filtering approach, handle low resolution images and due to compression not possible to restore complete data. Detecting forgery from large volume of data is a challenging task [16]. Authors in [6], introduced new method of forgery detection based median filtering to handle compressed data images, in their model they used regression techniques and Markov chain model, these two methods in turn use dimensionality reduction while detecting image forgery.

Image forgery can be detected using forensic techniques. It can be useful in many legal service oriented applications, and in turn these techniques resolve many criminal cases. In court of criminal laws separating original and duplicate images is more important task. Taxonomy of various forensic techniques is listed in Figure 1.

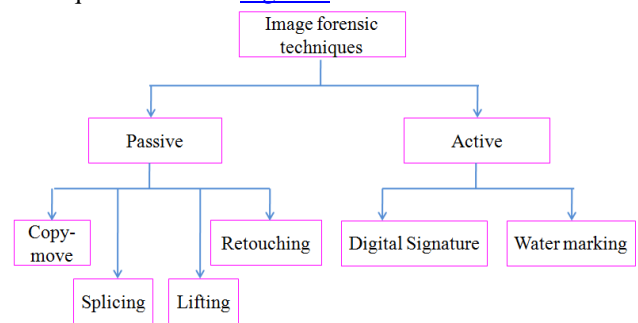


Figure 1: Classification of Image Forensic Techniques

Now a day's number of tools are available to change original images, with help of these tools one sub area of an images easily tampering with other, original and duplicate images are at most similar, and distinguish both images are one of the challenging task. To identify image forgery two active approaches are used is digital signature and other one is water making.

In digital signature (DS) image are partitioned in to 16×16 parts, for each sub area apply discrete wavelet transform, trace out DS, calculate hash code of traced out DS, compared received and calculated hash code, and with that identify forgery if any one of the signature mismatch. In water making technique check sum is added at LSB position, add length of input image, also calculate correlation coefficient of water maker images, and water marked images are not visible scanners or humans.

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To identify image forgery three passive approaches are used and they are retouching, splitting and copy move method. Copy move forgery is one of the special categories of image forgery in which some area of an image is copied and replicated in the same image with an intention to hide valuable feature [10] of the image. In retouching forgery method image features are changed by rotating original image, resizing by changing resolution, and modify original image by stretch. In image splicing mostly used high contrast to modify edges of an image, apply image smoothing technique to smooth sub parts of an image or its corners.

II. RELATED WORK

In the market new updated editing tools and technologies are available to edit images and with help of these tools images are easily forged. Authors in [4] introduced forgery detection based on “copy-move” concept. They used many feature of images to detect forgery with high speed and also used exponential transform technique. Authors in [4] first partitioned image into small size blocks (b_1, b_2, \dots, b_n) using segmentation technique, and all these blocks are divided into two types. One is smooth and second is texture, images are compared for matching, and identify false or fake or forgery detection using proposed method.

In image processing area detecting forgery images under cut and paste mechanism is more complex than other image editing techniques. Authors in [5] proposed texture based method to detect forgery images, they divide images into different local regions, for each local region calculated median and entropy values, they used morphology technique to identify features, and they used new used kind of support vector machine to improve performance of forgery detection algorithm. In their research they have taken images of different types, they used window size of 3×3 and 5×5 , all images are compressed before applying it to their method, and authors in [5] proved that their method will detect forgery even they used cut-paste mechanism to edit images.

Authors in [7] used median filtering to handle images of compressed, it is more important in image forgery detections or detection of fake video. Authors in [7] used machine learning technique (CNN) to handle input image, it is one of the self train and learning method. Their method resolve over filtering process and their method detect forgery even in compressed images. Image sharpening is a technique to enhance image visuals, if images are edited it is not possible to authenticate, the main goal of image forensic is to identify editing images, and to resolve this issue authors in [8] proposed new technique for image forgery based on image globalization. To analyze image pixels authors in [8] used more than ten different sets to extract features of input image, these features are useful to detect image forgery and performance of their method is evaluated with various types of data sets.

Due to limited availability of resources majority of deep learning techniques accept input images of small size and also these images are used resizing function to resize. Now a day images with high resolutions are available when applying resizing its quality will impact and it will degrade overall performance of a deep learning algorithms. Authors in [2] proposed CNN based forgery detection of images, processed

images as patch, all high resolution are accept by their method, used check point method to train data samples, all parameters are restricted to use less memory space and take decisions with high resolution images whether it is forgery or original.

Morphology is more popular non linear functions, and widely used in many applications to discover features, to enhance images, and also useful to remove noise from images. Morphology based filtering approach useful to detect artifacts from edited images, and this method is used in image forensics. Authors in [9] extended deterministic model to detect forgery and used SVM for classification to extract image features. Authors in [12] conducted survey on image forgery problem, identification of resizing of original image, taken images without compressed, analyzed background of images, if the data is multimedia then detection of fake videos takes more time and they identify resizing of images using CNN method.

Now a day's information posted in the form of blogs, articles and libraries in various social media networks. Modern tools or software are available to edit images and it is difficult to identify original and duplicated or edited images. Earlier researchers on forgery detection proposed methods mostly based on handcraft mechanisms, problems facing by these methods are they will identify image forgery on a particular domain, and they are incapable to detect on all categories of images. Authors in [16] conducted detailed survey on forgery using deep learning technique with various types of image datasets. Forensics is now mostly used in many criminal cases to identify fake images to solve cases. Authors in [11] proposed a technique to detect tampered image, verifies authentication of images, and calculate hash code to detect image tampering.

III. PROPOSED METHOD

In this research paper we proposed new forgery detection technique with estimation of noise on various scale of input image affect of noise in input image occur several times, frequency of images are also changed due to noise, noise signal correlated with original input images and in compressed images quantization level frequency also changed due to noise. Overall idea of proposed forgery detection method is shown in [Figure 2](#).

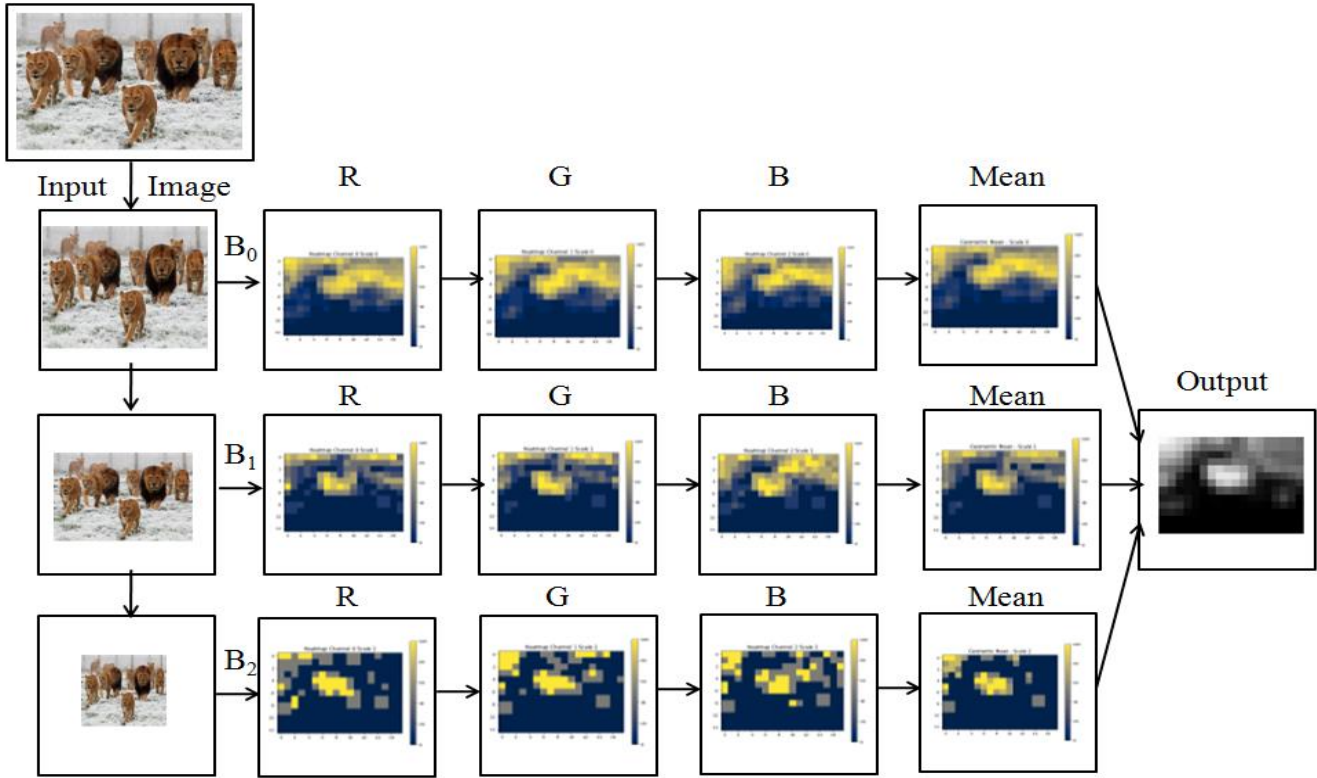


Figure 2: Forgery Detection Process

In this situation calculation of noise at local and global level is more important in forgery detection method. For example input image (I) has C number of colors on image channels. We partition input image into M X N blocks, resized blocks are proceed further, image colors are also taken into consideration, each block noise value is evaluated at local level and global level. For each color channel of input image estimate local and global noise levels are estimated and compared using binning method. Also measured heat map of each block and each color channel of input image and all these values are stored in bins. Finally from all noise values calculate average mean value of noise. Algorithm is proposed to detect image forgery is as follows.

Algorithm:

Step 1: Select image which is suspect of forgery. Sample input image is shown in [Figure 3](#).



Figure 3: Suspected Input Sample Image

Step 2: Fix block size and partition input image into smaller blocks based on block size.

$I = \text{Input Image consist of } (M_x, M_y) \text{ Blocks}$

$C = \text{Color Channel}$

$$H_x = \left\lfloor \frac{M_x}{B \times S} \right\rfloor - 1, \text{Number of Horizontal Blocks.}$$

$$H_y = \left\lfloor \frac{M_y}{B \times S} \right\rfloor - 1, \text{Number of Vertical Blocks.}$$

Step 3: For each small block analyze color level, analyze histogram level at image boundaries, observe colors of block image and also calculate noise level of each smaller block and as well as global noise level of suspected image of forgery.

Step 4: Calculate heat map of each color channel of input block image and it is scaled to (0,255). To get output at the end merge all heat maps of individual smaller blocks.

Step 5: Noise level of image at global level is minimum then image is not tampered and if noise is high then decide image is tampered.

IV. PERFORMANCE EVALUATION

We executed proposed model, results are evaluated using scaling techniques, results are compared by varying one scale (B0), double scale (B1) and triple scale (B2), and also results noise values of blocks of proposed method compared with existing methods like splicing, retouching, and so on. To analyse performance of proposed method CG1050 dataset is used, these database has at most 1050 images, all these images are classified into four categories, images of this dataset are taken in different locations, all these images are captured with high resolution, all images are stored in two different forms one is colour and other one is gray [13], actually compressed version of images are stored in dataset, and for all these images original and as well as duplicate images are stored in the dataset.

Forgery detection is a binary classification technique, number of class label present is two (positive (yes or forged), negative (no or not forged), performance of a proposed method is measured with help of confusion matrix [14] [15]. Confusion matrix format to evaluate performance of proposed algorithm is shown in Table 1.

Table 1: Confusion Matrix

		Predicted	
		P	N
Actual	P	True +ve	False -ve
	N	False +ve	True -ve

Let 'a' is a pixel, H is heat map value, M is mask of forgery, and four possible combination of confusion matrix values are calculated using following equations.

$$TP_I = \sum_a H(a) * M(a)$$

$$TN_I = \sum_a (1 - H(a)) * (1 - M(a))$$

$$FN_I = \sum_a H(a) * (1 - M(a))$$

$$FP_I = \sum_a (1 - H(a)) * M(a)$$

Let 'r' correlation coefficient, 'b' true positive, 'c' true negative, 'd' false negative, 'e' false positive, connection by Joining (J) and F1 is a score. These three performance evaluation metrics are used to proposed forgery method and are defined as follows.

$$r = \frac{b * c - d * e}{\sqrt{(b + e) * (b + d) * (c + e) * (c + d)}}$$

$$F_1 = \frac{2 * b}{2 * b + d + e}$$

$$J = \frac{b}{b + d + e}$$

Input image is scaled to one scale (O), double scale (D), Triple scale (T), correlation coefficient value (r) is measured using different forgery detection methods (Splicing (S), Coloring (CR), Retouching (R), and Copy move (CM)), and results are shown in Figure 4.

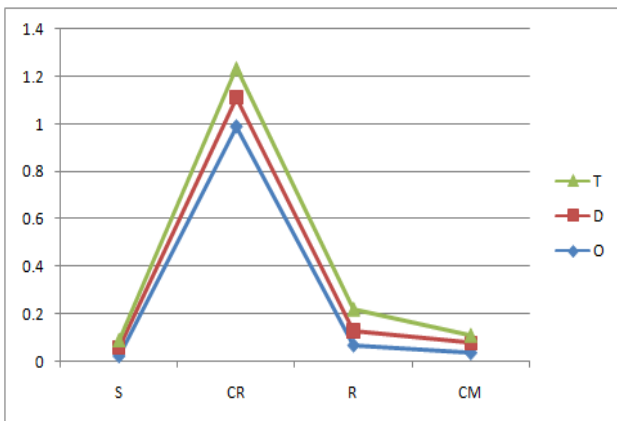


Figure 4: Forgery methods Vs. Correlation coefficient

Input image is scaled to one scale (O), double scale (D), Triple scale (T), correlation connection by Joining (J) is measured using different forgery detection methods (Splicing

(S), Coloring (CR), Retouching (R), and Copy move (CM)), and results are shown in Figure 5.

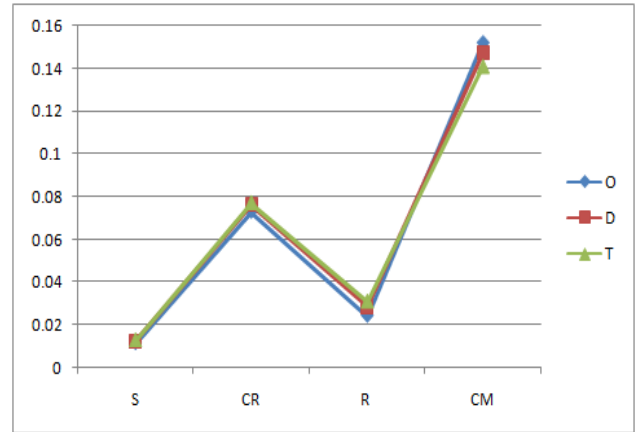


Figure 5: Forgery methods Vs. connection by Joining

Input image is scaled to one scale (O), double scale (D), Triple scale (T), correlation F1 Score is measured using different forgery detection methods (Splicing (S), Coloring (CR), Retouching (R), and Copy move (CM)), and results are shown in Figure 6.

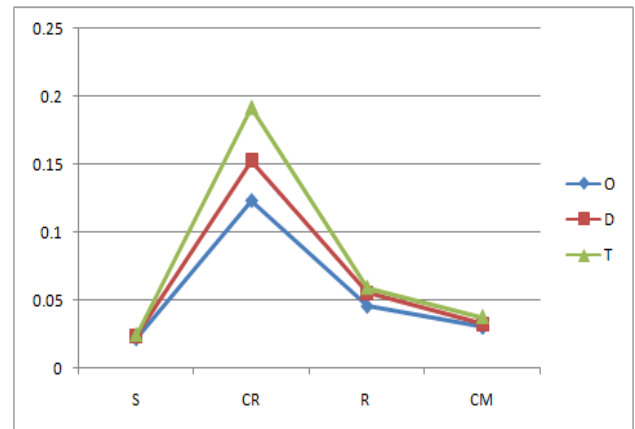


Figure 6: Forgery methods Vs. F1 Score

V. CONCLUSION

With latest development tools and technology, an image can be easily editable, untrained peoples may not in a position to differentiate original and forged images, many image processing tools are available to detect such a forgery images, all these tools are giving clues to detect forgery, a window sized boxes will scan entire images based on clues, and all these image forgery detection techniques are suffering with boundary problems. In this paper we proposed new forgery detection method to detect tampered images. Images are fitted into three different scales, color of the images and heat map of each channel are calculated, noise level are measured local areas and global areas of images, and from these values easily identify tampered images. Performance of proposed method is analyzed by using correlation coefficient, connection by Joining, and F1 Score.

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