

Trace Disclosure Analysis Using Quantization Noise Scheme in High-Quality JPEG Compression

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Abstract— Nowadays, a huge number of images are available to find the image tampering in a forensic environment. Due to its low-cost nature, the image acquisition and dissemination of digital images makes a great demand in conveying the information. To detect whether an image is JPEG compressed or not in forensic environment is an important issue. The nature of transform coding is lossy that makes the JPEG system to analysis on its traces of the images. Existing method fails on high-quality compressed image such as those with a quantization table containing mostly quantization steps being ones. The earlier technique doesn't deal on the high-quality JPEG compressed images. In this paper, we focused on high-quality JPEG compressed images. We introduced a novel scheme known as, Forward Quantization Noise (FQN) which investigated on the Trace Disclosure Analysis (TDA) of the compressed images. This scheme worked on the analysis of noises in images over multiple passes. We derived that the decompressed image exhibits lower variance than the uncompressed image. We introduced a detection algorithm that detects the high-quality decompressed images. From various sources of images, we experimented that our scheme outperforms better than the previous schemes in terms of robustness and chroma subsampling. Our proposed detection algorithm can efficiently execute in real-time applications such as Internet Image Classification, fraud detection, etc.

Index Terms—JPEG Compression, Forward Quantization noise (FQN), Fraud Detection.

I. INTRODUCTION

There are several types image compression standards exists, including lossy and lossless. These exist because several kinds of requirements needed on image quality, storage, and transmission. In these, JPEG is a popular lossy compression standard. The main interest of image forensic experts on JPEG compressed images from unknown sources because to detect the image forgeries [2], [3] and processing history of images. There are some reported results on

identifying whether an image is compressed [4],[5], or uncompressed, whether compressed an image twice or thrice [6]-[10] with shifted grid position

[11]- [13], and on estimating the JPEG quantization steps [5] and quantization table [14], [15]-[17].

In this paper, we concentrate on whether an image is truly in uncompressed form or has been previously JPEG compressed. Finding such historical records may help to answer forensics related questions i.e., where the image coming from, whether it is true image or forgery image. This solution may help to detect image forgeries created by replacing some part of image with another part of image with different compression historical records. The mismatch of historical records identify image forgeries. This JPEG identification problem [3], [4] may also estimate JPEG quantization steps. So, the forensic experts save time filtering out uncompressed images and performing estimate only on decompressed images [17]-[19].

High quality compressed images similar to uncompressed images due to their nearly lossless nature. So, this high quality compressed images used with uncompressed images for creating forgeries. Current forgery detectors [4], [5] are not capable are not detect high quality compressed images even in the anti-forensics detectors. So, this is the main problem to identify high quality compressed images when they decompressed and saved in uncompressed form.

Traces of JPEG compression might be found in image intensity domain. If we quantize high frequency discrete cosine transform coefficients with large quantization steps produces ringing effects when we saved in decompressed form. It produces blocky artifacts [20], [21] in case of heavy compression. In existing method Fan et al [4] describes an image is JPEG compressed based on pixel differences within the 8×8 blocks is similar to those across the blocks. This method is effective in detecting heavy compression of blocky artifacts. In case of an image is small size and high quality compression this method is not reliable.

Traces of JPEG compression may also be found in the histogram of DCT coefficients. Luo et al [5] noted that JPEG compression reduces the amount of DCT coefficients within the range of $[-1, 1]$. After an image is JPEG compressed the DCT coefficients within the range of $[-1, 1]$ are small. The amount of DCT coefficients in the range of $[-2, 2]$ is also found. When the statistics of a test image exceed the threshold then it is classified as an uncompressed image, otherwise it is classified as a previously JPEG compressed image. This method has few drawbacks. The first one is, it uses DCT coefficients that are close to 0. So, the information is not fully utilized. The second one is, it uses quantization steps above 2. So, this method fails to detect high quality compressed images because it contains quantization steps around one.

During this paper, we define a noise called as forward quantization noise. We develop a simple algorithm to detect whether an image is JPEG compressed based on the variance of forward quantization noise. This method fully utilizes noise information on DCT coefficients. So it is neither limited to an image of large size nor quantization step being under 2. Finally we show that our method is good compared to previous methods. This paper is organized as follows. Section II describes theoretical work on multi-cycle JPEG compression. From this we show how the variance is employed on forward quantization noise to detect JPEG compressed images in section III. In section IV we describe results. The paper is concluded in section V.

II. JPEG QUANTIZATION NOISE ANALYSIS

A JPEG compression cycle [22], [23] consists of an encoding phase and a decoding phase. In the encoding phase, data loss occurs due to the quantization of DCT coefficients. The decoding phase is the reverse of the encoding phase. The truncation and integer rounding operation occur when JPEG coefficients are restored in the image intensity domain. In a recent study [22], we represent a work for analyzing multi-cycle JPEG compression based on a complete JPEG compression model compared with a simplified model [5], [23]. This study focused on information losses in JPEG compression, which are quantization noise and rounding noise. The truncation error is ignored because it has less impact on this model. The distribution of two noises is derived. In this section, we summarize some results from the work and introduce some conventional notions.

A. Notations

In this total paper, the noises introduced during JPEG compression are in lower case symbols and DCT coefficients are in upper case symbols. The DCT coefficients in 8×8 grid are numbered from 1 to 64. The first coefficient ($u=1$) is a DC

coefficient due to its low-pass property and it is a mean of all pixel values in an 8×8 block. Remaining coefficients ($u=2 \dots 64$) are called as AC Coefficients due to their high-pass property. The index u is also used to indicate corresponding noise in the DCT domain. The index m is used to indicate pixels in the spatial domain and their noise is indexed from 1 to 64.

The processing diagram for multi-cycle JPEG compression is shown in the left part of Fig. 1. $X^{(k)}$ and $\hat{X}^{(k)}$ are used to denote floating point images in JPEG encoding and decoding phases respectively in the k -th JPEG compression cycle. $Y^{(k)}$ and $\hat{Y}^{(k)}$ are used to describe un-quantized DCT coefficients in the encoding phase and de-quantized coefficients in the decoding phase. $I^{(k)}$ and $\hat{I}^{(k)}$ are used to represent image integer representations. NIL means no processing step. $W^{(k)}$ is used to represent quantized DCT coefficients. The logical diagram for multi-cycle JPEG compression is shown on the right side of Fig. 1, obtained by dropping the NIL operation. In the logical diagram, rounding noise is denoted by $x^{(k \rightarrow k+1)}$ and $y^{(k)}$ denotes quantization noise, and we also define two auxiliary noises $x^{(k)}$ in the spatial domain, $y^{(k \rightarrow k+1)}$ in the DCT domain.

B. Quantization Noise

The quantization noise is defined as the information loss due to the JPEG quantization process, i.e.,

$$y = Y - \hat{Y} = Y - \left[\frac{Y}{q} \right] q, q \in \mathbb{N} \quad (1)$$

Where q is the quantization step and $[\cdot]$ represents the integer rounding operation.

C. General Quantization Noise Distribution

In general, the distribution for quantization noise is given as

$$f_y(s) = \sum_{k=-\infty}^{\infty} f_Y(kq + s), s \in \left[-\frac{q}{2}, \frac{q}{2}\right), k \in \mathbb{Z} \quad (2)$$

Where q is the quantization step, f_Y and f_y are respectively the distributions for Y and y .

D. Specific Quantization Noise Distribution

In previous work we found that the quantization noise of the first round compression cycle is different from the second round compression cycle.

The distribution for quantization noise for the first round cycle is given as

$$y_u^{(1)} \sim \begin{cases} U\left(-\frac{q_1^{(1)}}{2}, \frac{q_1^{(1)}}{2}\right), & u = 1 \\ Q^L(\lambda_{y_u^{(1)}}, q_u^{(1)}), & u \in \{2, 3, \dots, 64\} \end{cases} \quad (3)$$

Where U represents uniform distribution and Q^L represents quantized-Laplacian distribution. $q_u^{(1)}$ is the quantization step of the u -th frequency in the first compression cycle.

For all DCT coefficients of u, the second-cycle quantization noise distribution is

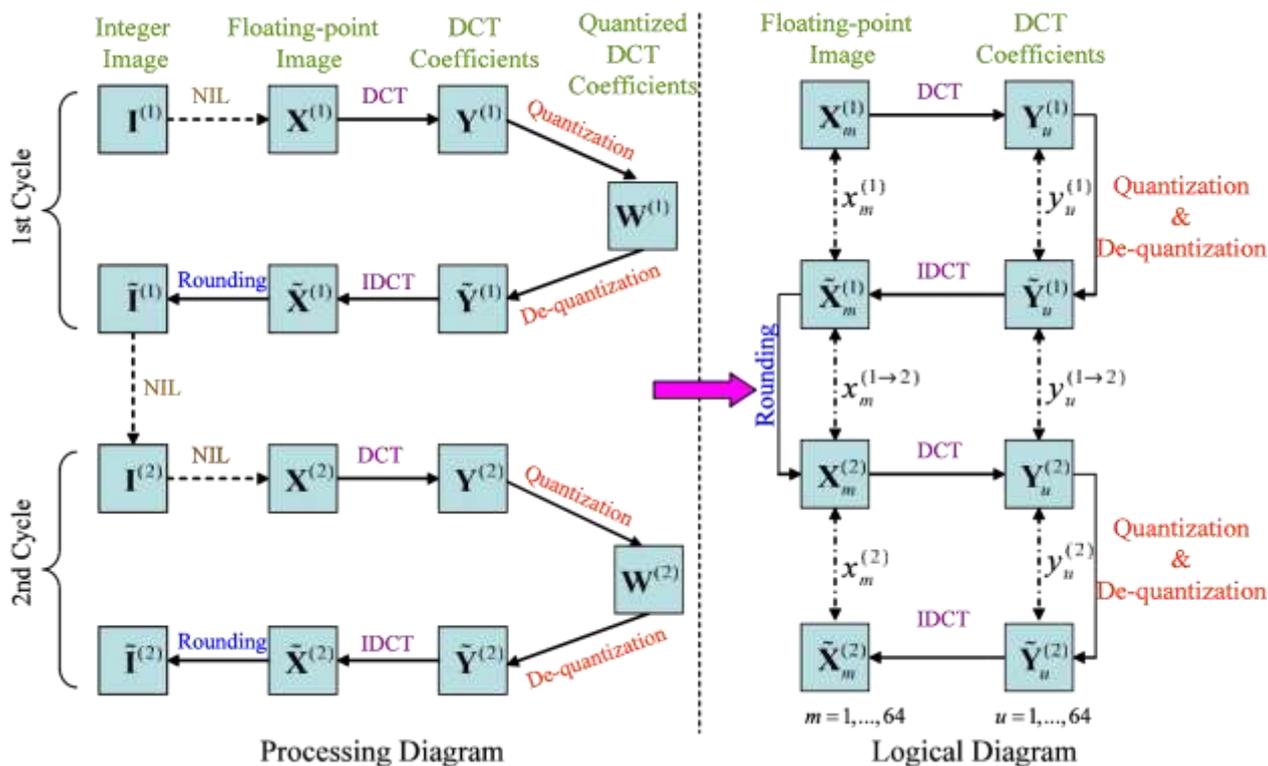


Fig 1.Processing steps for multi-cycle JPEG compression

$$y_u^{(2)} \sim \begin{cases} Q^N(\sigma_{y_u^{(1-2)}}^2, 1), & q_u^{(2)} = 1 \\ N(0, \sigma_{y_u^{(1-2)}}^2), & q_u^{(2)} \geq 2 \text{ and } (\frac{q_u^{(1)}}{q_u^{(2)}}) \in \mathbb{N} \\ f_y \text{ as in Equation (2), otherwise} \end{cases} \quad (4)$$

Where Q^N is quantized-Gaussian distribution. The distribution of $y_u^{(2)}$ may depend on the variance of the auxiliary noise $y_u^{(1-2)}$.

III. IDENTIFICATION OF DECOMPRESSED JPEG IMAGES BASED ON QUANTIZATION NOISE ANALYSIS

In the following section we discuss forward quantization noise, and gives it relation to the quantization noise. Then we give upper bound for noise variance of uncompressed and compressed images. Finally, we give simple algorithm to differentiate decompressed images from uncompressed images.

A. Forward Quantization Noise

If we give an uncompressed image in JPEG encoding phase, we can get quantization noise of that cycle. Suppose if we give a compressed image, but stored in uncompressed we did not quantization noise of that cycle, but we can find quantization noise of next cycle .So, finally we say that quantization noise

available for current available cycle as forward quantization noise. Quantization noise is function of quantization step. From this work the simplest form of quantization noise with quantization step 1 i.e.

$$z=Y-[Y], \quad (5)$$

Where Y is DCT coefficient

For an uncompressed image the forward quantization noise is first compression cycle quantization noise with quantization step one, i.e., $q_u^{(1)} = 1, u \in (1,2,\dots,64)$. For a compressed image the forward quantization noise is second cycle quantization noise with $q_u^{(2)} = 1, u \in (1,2,\dots,64)$.

B. Noise variance for compressed images

The upper bound for the variance of forward quantization noise of an uncompressed image:

$$\sigma_z^2 = \sigma_{y^{(2)}}^2 \leq C_0 \quad (6)$$

where $C_0 = 0.0833$

C. Noise Variance for Images With Prior JPEG Compression

The upper bound for the variance of forward quantization noise of an image with prior JPEG compression:

$$\sigma_z^{(2)} = \sigma_{y^{(2)}}^2 \leq \begin{cases} C_2, & \text{if } q_u^{(1)} = 1, \forall u, \\ C_1, & \text{otherwise} \end{cases} \quad (7)$$

By combining (6) and (7) we have get following result for forward quantization noise. Given a test image I, the variance of forward quantization noise z with q=1 is given by

$$\sigma_z^{(2)} \leq \begin{cases} C_0, & \text{if } I \text{ is uncompressed,} \\ C_1, & \text{if } I \text{ was compressed once.} \end{cases} \quad (8)$$

Where $C_0=0.0833$, $C_1=0.0638$, $C_2=0.0548$

D. Identification of Decompressed JPEG Images

The above result on noise variance is derived theoretically. The distribution of practical samples may vary from theoretical result because of finite sample size. The variance of the practical samples is denoted by $\hat{\sigma}_z^{(2)}$, may slightly exceeds upper bound, i.e., C_1 or C_0

$$I = \begin{cases} \text{uncompressed,} & \hat{\sigma}_z^{(2)} > T, \\ \text{decompressed,} & \hat{\sigma}_z^{(2)} \leq T, \end{cases} \quad (9)$$

Where T is a predefined threshold which is in between C_1 and C_0

IV. EXPERIMENTEL RESULTS.

To evaluate the results 256×256 images are used .First we take an image and then it compressed with quality factor 95. Fig 2 gives JPEG compressed and decompressed image. In this results only evaluated parameter is variance. The variance shown in fig 4. For an uncompressed image the variance is below 0.0833. If an image is compressed the variance is below 0.0638 is shown in fig 4. An example for tampering is shown in fig 3.

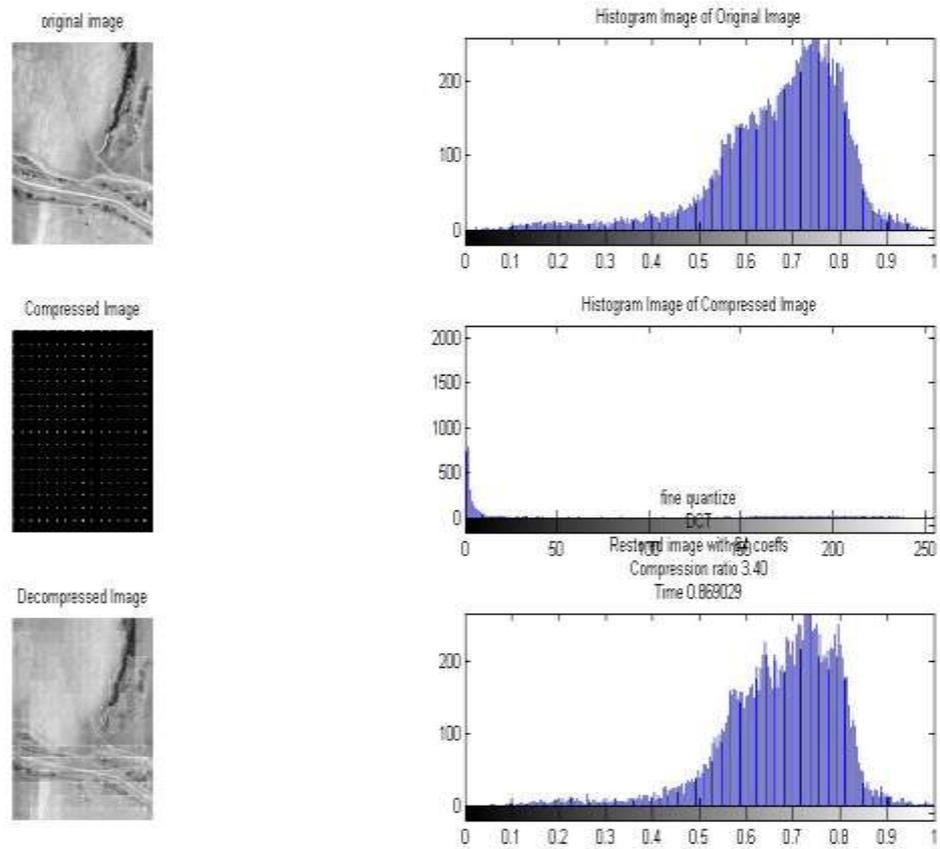


Fig 2. JPEG uncompressed and decompressed image

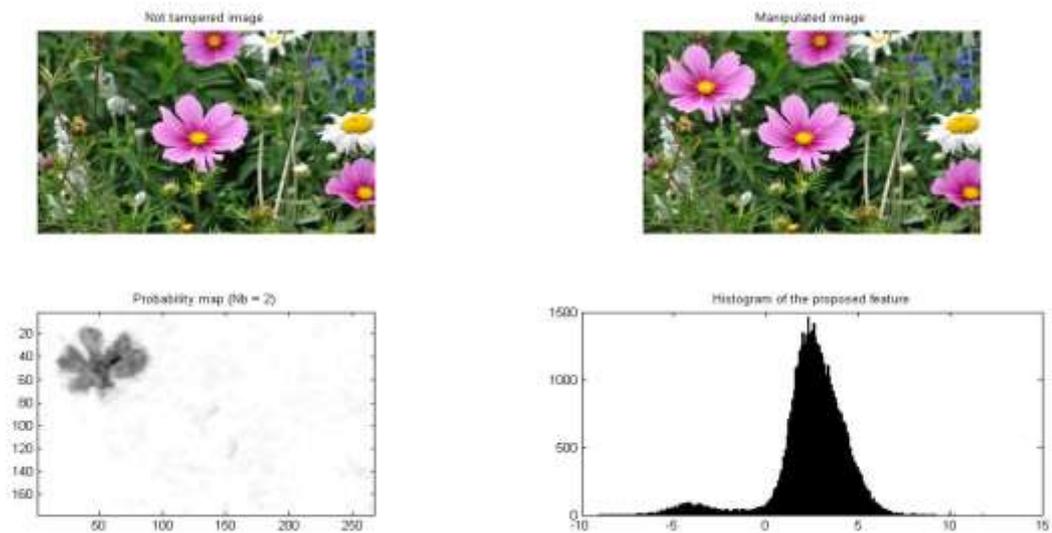


Fig 3. An example for tampering.

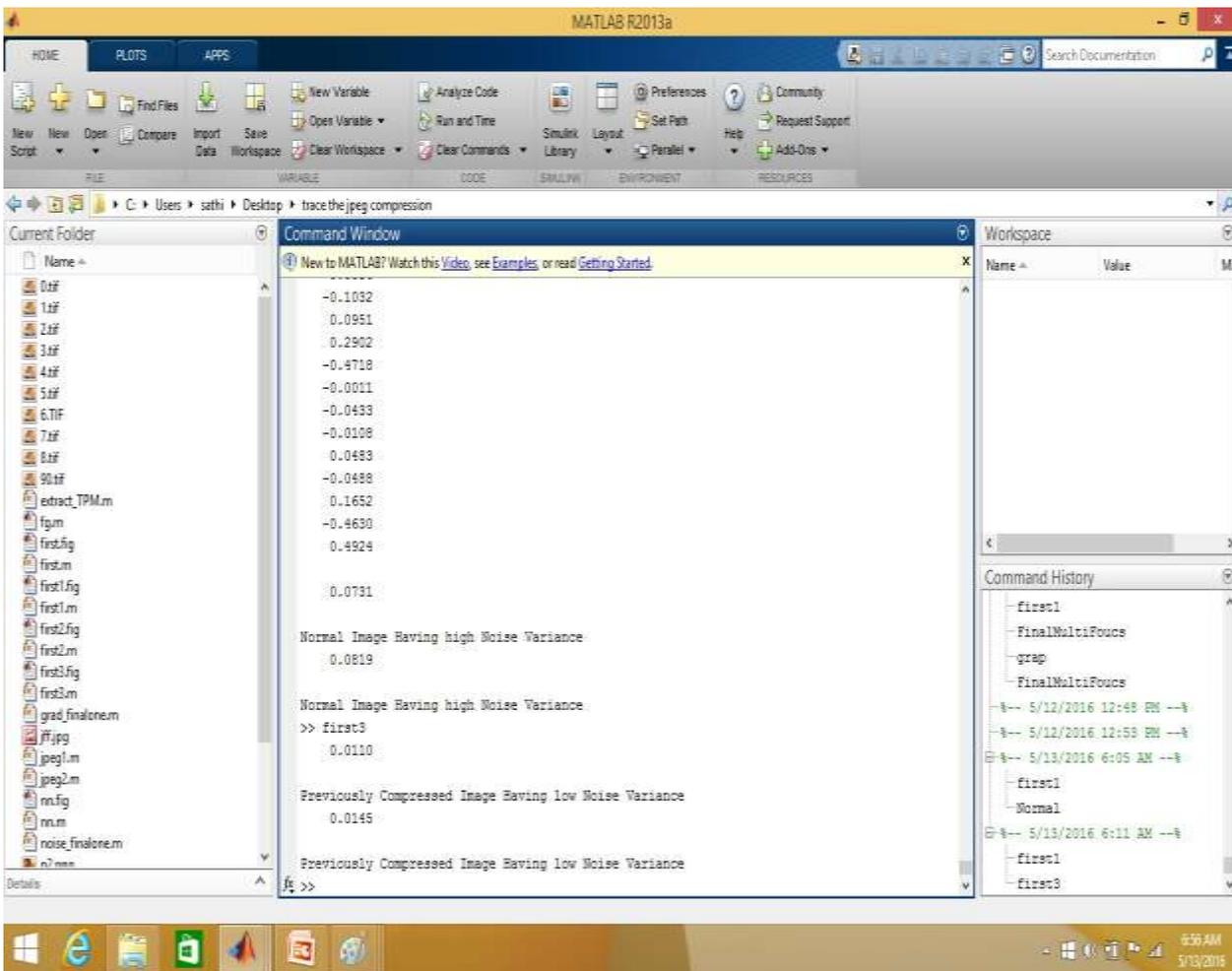


Fig 4. Comparing noise variance.

V. CONCLUSION AND FUTURE SCOPE

In this paper, we propose a method to reveal the traces of JPEG compression. The proposed method based on the variance of forward quantization noise. We derived decompressed JPEG image is lower noise variance than the uncompressed image. We applied this method challenges posed by high quality JPEG compression identification. The proposed can be applied to forgery detection. Our future studies will be on trying to extend the noise analysis to other forensic task, i.e., identifying the resized decompressed JPEG images such as images presented in IEEE IFS.

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