

A study of Content Based Watermarked Image Retrieval in Clouds with Privacy Protection

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Abstract

Content Based image retrieval is one of the major substantial research area for retrieving visual information on the cloud. Image retrieval may undergo irregularity between the specified and retrieved content. Using Content-Based image retrieval system, it is possible to retrieve only the relevant images. The image owners outsource their images to cloud, but before outsourcing, there may be a chance of image leakages and they may be faked. So the retrieved faked images from cloud may be dangerous in case of sensitive fields. So here in CBIR systems, watermarking technique is combined with encryption. In this paper a Content Based Image retrieval systems are used for querying a very large watermarked image databases. This makes the image leakages as well as retrieving faked images to get reduced and provide privacy for preserving them. This paper demonstrates the Secure and Robust Digital Image Watermarking using Coefficient Differencing (DCT) and Chaotic Encryption.

Keywords: *CBIR, watermarking, DCT, chaotic encryption*

Due to recent development in technology, there is an increase in the usage of digital cameras, smartphone, and Internet. The shared and stored multimedia data are growing, and to search or to retrieve a relevant image from an archive is a challenging research problem [1–3]. The fundamental need of any image retrieval model is to search and arrange the images that are in a visual semantic relationship with the query given by the user. Most of the search engines on the Internet retrieve the images on the basis of text-based approaches that require captions as input [4–6]. The user submits a query by entering some text or keywords that are matched with the keywords that are placed in the archive. The output is generated on the basis of matching in keywords, and this process can retrieve the images that are not relevant. Content-based image retrieval (CBIR) is a framework that can overcome the abovementioned problems as it is based on the visual analysis of contents that are part of the query image. By outsourcing CBIR services to the cloud server, the data owner is relieved from maintaining local image database and interacting with database users online. Despite the tremendous benefits, image privacy becomes the main concern with CBIR outsourcing. This paper considers watermarking based content image retrieval. The watermarked image extracted from the recognized image is matched with the test image

1. INTRODUCTION

watermark for authentication. However, the data when outsourced to the cloud might be corrupted or lost due to the inevitable software bugs, hardware faults and human errors in the cloud [1]. It is aimed to develop a system, that should enable to sense about that faked images. We hope that Content-based image retrieval (CBIR) and watermarking based image security is the appropriate solution.

A. Content Based Image Retrieval

The central issue in CBIR is to identify a set of salient image features for indexing and similarity evaluation. Color, shape, texture and spatial relationships among segmented objects are typical features employed for image indexing. Some researches combine two or more of these features to improve retrieval performance. The main goal in CBIR system is searching and finding similar images based on their content. To accomplish this, the content should first be described in an efficient way, e.g. the so-called indexing or feature extraction and binary signatures are formed and stored as the data. When the query image is given to the system, the system will extract image features for this query. It will compare these features with that of other images in a database. Relevant results will be displayed to the user. Fast and accurate retrievals among the data collections can be done according to the content description of the query image. There are many factors to consider in the design of a CBIR systems based on the domains and purposes, choice of right features, similarity measurement criteria, indexing mechanism, and query formulation technique. The most important factors in

the design process is the choice of suitable visual features and the methodologies to extract them from raw images, as it affects all other subsequent processes. By the nature of its task, the CBIR technology boils down to two intrinsic problems: (a) how to mathematically describe an image which can also be called as feature extraction. (b) how to assess the similarity between a pair of images based on their abstracted descriptions. Which can also be called as matching. [2] In typical content-based image retrieval system the visual contents of the images in the database are extracted and described by multi-dimensional feature vectors. The feature vectors of the images in the database form a feature database. To retrieve images, users provide the retrieval system with example images or sketched figures. The feature vector of this query image is extracted. Then the extracted feature vector of query image is compared with the feature vector in the database. As a result of the query, similar images are retrieved according to their similarities between the feature vectors of the query example or sketch and those of the images in the database. In this paper I have proposed a content-based image retrieval method based on an efficient combination of color, texture and neural network based ICM (Intersecting Cortical Model) features. Both color and texture features of images are extracted and stored as feature vectors in a database. Then from those features image signatures are derived which basically provide the no. of white pixels in those 25 images and saved in the database for retrieval purposes. During the retrieval process, the color, texture and ICM feature vector of the query image is computed and matched against those features in the database. For image

matching there are various methods such as Euclidean distance method, Canberra distance, Man Hattan distance and city block distance method are available for color and texture features and k-means method for comparing the ICM features. Here, in this paper city block distance have been used for distance matching, which gives better result for CBIR.

B. Digital Watermarking

It is important to outsource data to the cloud since the information technology is improved. The security level of data should be enhanced when the development also increases. There are several privacy issues against which the sensitive data is to be protected and there are several techniques proposed for securing the data. Here I have chosen digital watermarking technique. For securing the original data from illegal manipulation and distribution, a label, tag or information container is added within the multimedia data through watermarking technique. Visible or Invisible Watermarking is the two different types of watermarking approaches. Depending upon the type of application in which watermarking system is applied, the properties of watermarking are developed. The effective detection of watermark is very important. It is not possible to fulfill the goal of watermarking when the watermark is detected. The watermarking approach considers the host signal quality property as an important one. The host signal is known to be affected by the watermark is embedded within watermarking. Thus, least numbers of changes must be done on the host signal. Also, when the watermark is not visible in the watermarking system, the host signal must be imperceptible. Watermark Size helps in performing

owner identification. The host signal is provided with security. It is very important to use this watermark while transmitting the data. Within all watermarking systems, robustness plays a very important role. The watermark is degraded at the time of transmission. The transmitted data may be corrupted. Thus, to ensure the system toleration, the robustness of watermark is necessary. The user's data is ensured to be safe by authenticating it.

Contribution:

The contribution of this paper can be summarized as follows:

- (i) Before outsourcing the image, watermark will be embedded with the original image using DCT based digital watermark algorithm and watermark will be encrypted using Chaotic encryption.
- (ii) After embedding, IDCT is applied and a robust watermarking image will be generated. Then the Watermarked image will be outsourced to the cloud.
- (iii) Based on query image, color and texture features retrieves better sets of similar images. The images retrieved will have similarity measure values between 0.0 and 1.0.

2. LITERATURE REVIEW

Abdulaziz Shehab, et.al (2018) proposed a novel approach for medical applications to provide image authentication and self-recovery. The image tampering is localized and the original image is recovered by proposing this new fragile watermarking based approach [3]. Further, the transformation of original image is identified by applying SVD with the help of inserting the traces of block wise SVD into the LSB of image pixels. For surviving the vector

quantization attack, block authentication and self-recovery bits are the two authentication bits that are applied. The Arnold transformation is applied for recovering the original image even when the tampering rate is higher for determining the insertion of self-recovery bits. There are several attacks in the presence of which the proposed mechanism is tested. It is seen through the experimental results achieved that the tamper location accuracy and PSNR of self-recovered image are enhanced to a higher level as compared to the existing approaches.

Zigang Chen, et.al (2018) proposed a new technique such that the image content can be protected and authenticated by applying General-NMF (General non-negative matrix factorization) based digital watermarking technique [4]. For the factorization of a matrix into two factor matrices, a general-NMF technique is proposed initially. There is no need to meet the dimension matching constraints in this proposed mechanism. Further, on the basis of proposed General-NMF, a digital watermarking mechanism is proposed here. Here, the keys applied for digital watermarking include the generator factor of random matrix and n . Several attacks and tampering are resisted effectively

by the proposed approach as per the simulation results achieved. It is also possible to apply the proposed mechanism as digital watermarking approach depending upon the feature matrix and transform domain since certain parts of faces and semantic features of text can be learnt by

NMF. Shuai Liu, et.al (2017) proposed a fractal encoding method and DCT based digital watermarking algorithm in this paper. For making enhancement in

the existing DCT method, the fractal encoding and DCT methods are integrated by the proposed method such that the encryptions can be doubled [5]. In the form of first encryption, the fractal encoding is used to encode the image. Further, to perform second encryption, the encoded parameters are used within the DCT method. For encoding one private image with private scales, the fractal encoding method is applied initially. In the form of digital watermarking, the encoding parameters are applied. Further, DCT is used to reversibly add digital watermarking within the original image. Thus, along with the private encoding scales, it is possible for the author to extract private image from the carrier image. It is seen that in comparison to the traditional approaches, the performance of proposed method in terms of robustness and PSNR is better as per the simulation results achieved.

B. Jyothi et al. (2016) [6] focused on texture components that are integrated with user feedback to improve the retrieval accuracy of the BIR system. Some complex techniques such as intensity, gradient and edge mapping techniques were used to achieve their goal. If the user is dissatisfied with the images retrieved, the system provides an interface where the user can select the most relevant image for searching again. The system analyses the relevance feedback from the user and returns better results.

Roshi Choudhary et al. (2014) [7] have described an integrated CBIR method to extract both texture and color features. A technique called color moment (CM) extracts the color features from colored images and local binary pattern (LBP) extracts the texture features from grayscale images. A feature

vector is then formed by combining the color and texture features of the image. The Euclidean distance between the feature vectors of the database images and query image are computed and images in the database corresponding to the query that fall below the Euclidean Distance threshold value are returned.

Yu-Chen Wang et al. (2015) [8] They have proposed a method called biased discriminant analysis which is a subspace learning algorithm, with feature line embedding for enhancing performance in relevance feedback schemes by identifying the correlation between feature lines and samples. Relevance feedback is used for dimensionality reduction and feature line embedding is utilized to reduce the gap between abstract high-level semantic features and low-level features.

Zhi-Chun Huang et al. (2010) [9] have attempted to simplify the complexity involved in computation and improve the retrieval accuracy of relevant images by combining colour and texture features and allowing the user to specify the weights that must be assigned to these features while measuring the similarity between images. They have used HSV (Hue, Saturation, Value) colour space for colour features. Gabor texture descriptors were used for the texture feature.

3. PROPOSED WORK

A. Digital image Watermarking using DCT and Chaos Encryption

Among the various transform domain techniques, DCT is considered as an efficient digital watermarking algorithm due to its low cost of

hardware design. DCT of a digital image is considered in three ways. It is computed either on the whole image or is computed on the blocks of the image or only the DC coefficient of blocks of image are computed [10][15]. In all the ways embedding will be done by altering the DCT coefficients. For copyright protection, a watermark must resist on images, and is secure from image processing attacks. To achieve this embedding is done in low frequency coefficients. If the embedding is done on low frequency coefficients, then modification of these coefficients will make a low quality watermarked image. So for authentication purpose, embed the watermark in high-frequency coefficients.

To achieve robustness embed the watermark in midfrequency coefficients [15]. Another major concern is security of watermark in digital image watermarking techniques. For that a cryptographic method called Chaotic Based encryption is preferred here which holds privacy preservation [11]. Chaos based encryption algorithms, on the other hand, show highly efficient results due to their excellent characteristics viz. initial condition sensitiveness, periodicity, pseudorandom behavior, and ergodic nature. Due to these properties, the adoption of chaotic methods has been considered widely in recent years.

B. Chaos Encryption

A chaotic based encryption algorithm is an effective method for data encryption. Chaos signals possess the qualities of pseudo-randomness, irreversibility and dynamic behavior. The systems having chaotic nature possess high sensitivity to initial parameters. The output chaotic sequence is similar to white noise having random behavior with improved correlation and

complexity and is defined as reported in [12] and given by:

$$C_{n+1} = \mu \times C_n \times (1 - C_n) \quad (1)$$

where $0 < \mu < 4$ typically μ can be set to value 3.9 in order to achieve highest randomness and $0 < C_n < 1$ is the n th value generated from Eqn. 1. Different values of C_n could be obtained by varying the value of n from 0 to $L-1$. Here, L is the maximum number of chaotic values. By setting the initial values of μ and C_0 , we can get the required chaotic signal. As it offers the joint advantage of speed and security, the use of chaotic encryption has been shown to offer increased security [16]. The security of information can be increased by using various encryption. This encryption method, is two dimensional and works well in applications for encrypting images of type $N \times N$. The Arnold transformation is mathematically represented as

$$\begin{matrix} x^n \\ y^n \end{matrix} = \begin{vmatrix} 1 & 1 \\ 1 & 2 \end{vmatrix} \begin{matrix} x \\ y \end{matrix} \pmod{N} \quad (2)$$

C. System model

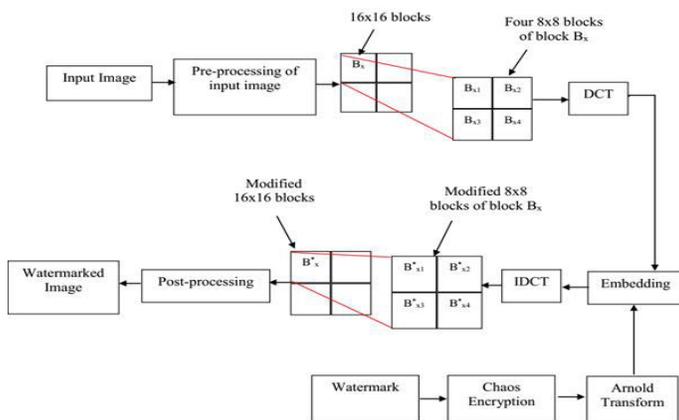
As shown in Figure 1, the input image 'I' is passed through the pre-processing unit which acts as a buffer for grayscale images and as a converter for color images. To carry out watermark embedding into the luminance part of the image the pre-processing unit

converts the input RGB image into YCbCr image, where

Fig:1 Digital watermark embedding using Coefficient differencing and chaotic encryption

Y stands for luminance information, Cb stands for chrominance blue information and Cr stand for chrominance red information of the image. The luminance part 'Y' is put forward as cover for the watermark because modification of this part of the image brings less noticeable changes to actual image compared to the chrominance information. On the other hand, if one wants to embed three watermarks into an RGB image, one in each plane, then the pre-processing unit extracts the RGB planes and then arranges all the three planes in a two-dimensional matrix so that each plane could be treated by the system as a $P \times Q$ grayscale plane, where P and Q respectively denote rows and columns of cover image. The resulting matrix values are brought in a range of -128 to 127 by subtracting 128 from the matrix.

After pre-processing the resultant matrix is divided into 16×16 blocks. For an input image of dimensions $P \times Q$, the number of blocks will be $P/16 \times Q/16$. Let an arbitrary block be represented by B_x . The block B_x is further divided into 8×8 blocks. The four 8×8 blocks of block B_x are represented by B_{x1} , B_{x2} , B_{x3} ; and B_{x4} as shown in Figure 1. Therefore, the total number of 8×8 blocks will be equal to $4 \times (P/16 \times Q/16)$. The total number of bits that could be inserted into a host image is equal to the total number of 8×8 blocks. The proposed technique utilizes the advantages of the



DCT coefficient correlation of adjacent blocks. Therefore, DCT of each block (8×8) is calculated. To embed a watermark bit, the difference between two preselected DCT coefficients of two neighboring blocks is calculated and is given as,

$$D = C_{xy}(i,j) - C_{xy+l}(k,l) \quad (3)$$

where $(i, j) \neq (k, l)$, gives the position of the selected coefficient within a sub-block and $1 \leq i, j, k, l \leq 8$. $x = 1, 2, 3, 4, \dots, P/16 \times Q/16$; whose value represents to which 16×16 pixel block the coefficient belongs while as $y = 1, 2, 3, 4$ (whose value represents the 8×8 DCT block to which the coefficient belongs). For the current work i, j, k and l are respectively taken as 3, 3, 3 and 2. For embedding the first watermark bit the difference between coefficient chosen from Block C_{x1} and a coefficient chosen from block C_{x2} is calculated. Similarly, to embed the second watermark bit, the coefficients from the block C_{x2} and the block C_{x3} are chosen for the difference, for embedding third bit in block B_x the coefficients from block C_{x3} and block C_{x4} are chosen and for embedding fourth bit the blocks C_{x4} and C_{x1} are taken for difference purpose. The difference 'D' is modulated according to the information bit to be embedded and the 'D' itself. The IDCT for every enhanced DCT block is calculated once the embedding process is completed. Further, post-processing operations are performed after IDCT. A final watermarked image is generated when the post-processing operations are completed. This watermarked image is outsourced to the cloud.

D. Image retrieval using various features

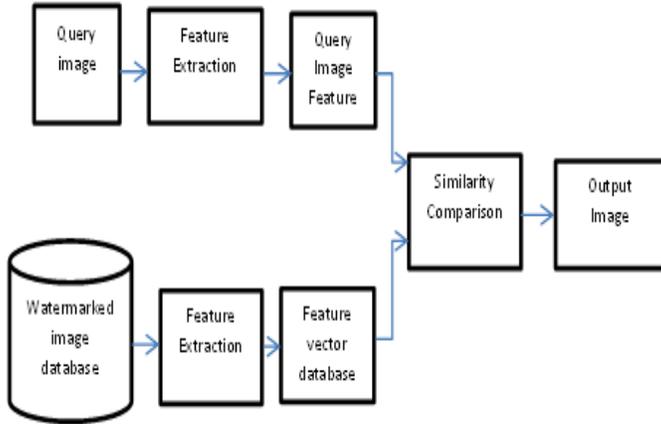
Most of the early studies on CBIR have used only

a single feature among various available features. However, it is hard to attain satisfactory retrieval results by using a single feature because, in general, an image contains various visual characteristics. Recently, active researches in image retrieval using a combination of color and texture features have been performed [1]. For an advanced CBIR, it is necessary to choose efficient features that are complementary to each other so as to yield an improved retrieval performance and to combine chosen features effectively without increase of feature vector dimension. When an RGB query image enters the retrieval system, it is first transformed into HSV color image. Then color feature is extracted and formed the color feature vector. Similarly the texture feature is extracted and formed the texture feature vector. After the color and texture feature vectors are extracted, the retrieval system combines these feature vectors, calculates the similarity between the combined feature vector of the query image and that of each target image in watermarked image database, and retrieves a given number of the most similar target images.

Color Features

Several methods for retrieving images on the basis of color similarity have been described in the literature [13], but most are variations on the same basic idea. Each image added to the collection is analysed to compute a colour histogram which shows the proportion of pixels of each colour within the image. The colour histogram for each image is then stored in

Fig:2 system architecture



the desired proportion of each colour (75% olive green and 25% red, for example), or submit an example image from which a colour histogram is calculated. Either way, the matching process then retrieves those images whose colour histograms match those of the query most closely. In this paper we have used HSV method for extracting the color feature and matched the result by using Newdist distance method.

Texture feature

In CBIR, texture features play a very important role in computer vision and pattern recognition, especially in describing the content of images. Texture features typically consist of contrast, uniformity, coarseness, and density. Importance of the texture feature is due to its presence in many real world images: for example, clouds, trees, bricks, hair, fabric etc., all of which have textural characteristics. Earlier methods [16],[17] for texture image retrieval suffer from two main drawbacks. They are either computationally expensive or retrieval accuracy is poor. Here we concentrate on the problem of finding good texture features for CBIR, which are efficient both in terms of accuracy and computational complexity. There are two main approaches for texture representations, statistical method and structural method. Structural methods

include morphological operators and adjacency graphs, describe texture through the identification of structural primitives and their arrangement in the image, for example, the presence of objects that are adjacent, perpendicular or parallel to one another. On the other hand, statistical methods identify texture through quantitative distributions of image intensity and include techniques such as Gabor, wavelet transforms, co-occurrence matrices, Fourier power spectra, Tamura feature etc. In this paper we have used Gabor technique for identifying the texture.

ICM Feature

ICM is a biologically inspired algorithm which takes as input an image and produces pulse images or binary images as output. The purpose or application of ICM is that each pulse image contains some important feature of an image like edges, shape, color and many more information that can be plotted and that plot can be used for image retrieval [ref1]. The three equations of ICM are:

$$F[n] = f F[n-1] + S + W(Y[n-1])$$

$$Y[n] = 1, \text{ if } F[n] > \Theta[n-1]$$

$$0, \text{ otherwise}$$

$$\Theta[n] = g \Theta[n-1] + h Y[n] \quad (4)$$

Where S is the input, F is the state of neuron, Y is the output, Θ is the threshold, f , g and h are constants, n is the number of iterations and $W(.)$ is the interneuron communication function. If the ICM iterates N times and outputs N pulse images Y , the signature will be a vector G with N elements. Each element is the number of white pixels in the corresponding pulse image:

$$G[n] = \sum Y[n], n=1,2,3,\dots,N \quad (5)$$

Since the pulse images from the ICM contain

different information from the image, such as form, edges and colors, the idea is that the information will be conserved in the signature.

Distance Metrics

Distance between two images has to be calculated to find if there is any match or not. Distance will help us in finding degree of matching for the entire data base. There are different distances available and here City Block distance method [19] is used. Distance between X and Y is calculated using equation $D\{(x_1, y_1), (x_2, y_2)\} = |x_1 - x_2| + |y_1 - y_2|$ (6)

This equation helps to find the city block distance between two points (x_1, y_1) and (x_2, y_2) . The distance between the x coordinates and y coordinates are found separately and the modulus value is added to get the distance.

Algorithm

1. For each image embed watermark using DCT.
2. The embedded image is stored in cloud which is called as watermarked image database.
3. If the user gives a query image, then do as follows,
4. Extract the Red, Green and Blue elements from an image and convert the R, G, B components to HSV color domain and store it in same image object.
6. Extract texture feature of the image using gabor filter and store texture feature vector in same image object.
8. Combine both the feature vectors using vector concatenation and store it in same image object [feature vector database].
9. Repeat Steps 4 to 8 for the query image.
10. For each image in the feature vector database, compute the distances using the user-selected similarity measure between the feature vector of

the query image and the corresponding feature vector of the image in the watermarked image database

11. End For

CONCLUSION

This paper describes the watermarking technique in general and also describe the watermarking using coefficient difference and chaotic encryption. From this technique, it is clear that encryption is applied on watermark, the watermark is secured and it provides high security as compare to the images after embedding. Many new techniques should be applied for better results in field of robustness and security. Here different methods are used for extracting color, texture as well as ICM features. And for similarity distance measurement city block distance method is used. Here it is concluded that DWT and ICM methods gives us better precision and performance. The main challenge in front of the CBIR system is time complexity in retrieving secured image.

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