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PROPOSED APPROACH FOR CONTENT BASED IMAGE RETRIEVAL IN CLOUD REPOSITORIES

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Abstract: Now days, Cloud computing it is been playing a crucial role in terms of data storing and reducing the overall cost to entrepreneurs. Storage requirements for visual data have increased in recent years, following the appearance of many interactive multimedia services and applications for mobile devices in personal and business scenarios. This was a key a determining factor for the adoption of cloud-based data outsourcing solutions. However, even the outsourcing of data storage in the cloud leads to new security challenges that must be carefully addressed. We propose a secure framework for the storage and recovery of the subcontracted privacy protection in large archives of shared images. Our proposal is based on IES-CBIR, a novel Encryption scheme of the image that presents image recovery properties based on content. The framework allows both encrypted storage and search using content-based image retrieval queries while preserving privacy against honest but curious cloud administrators. We have built a prototype of the proposed framework, formally analyzed and tested its safety properties, and experimentally assessed its performance and accuracy of recovery. Our results show that IES-CBIR is probably safe, allowing more efficient operations that the existing proposals, both in terms of complexity of time and space, and opens the way to new scenarios of practical application.

Keywords: Cloud computing, Data and computation outsourcing, encrypted data processing, searchable encryption content based image retrieval.

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I INTRODUCTION

Content-based image retrieval (CBIR) is also known as query by image content and content-based visual information retrieval is the use of computer vision to the image retrieval problem of searching for digital images in large databases. "Content-based" means that the search will analyze the actual contents of the image. The term 'content' in this context might refer colors, shapes, textures, or any other information that can be derived from the image itself. Without the ability to examine image content, searches must rely on metadata such as captions or keywords. Such metadata must be generated by a human and stored exactly each image in the database. An image retrieval system returns a set of images from a collection of images in the database to meet users demand with similarity assessment such as image content similarity, edge pattern similarity, color similarity etc. Image retrieval system offers an efficient way to access, browse, and retrieve a set of similar images in the real-time applications. As a result of recent advancements in digital storage technology, it is now possible to create large and extensive databases of digital imagery. These collections may contain millions of images and terabytes of data. For users to make the most of these databases effective, efficient methods of searching must be devised. Prior to automated indexing methods, image databases were indexed according to keywords that were both decided upon and entered by a human categorizer. Unfortunately, this practice comes with two very severe shortcomings. First, as a database becomes increasingly large the manpower required to index each image becomes less practical. Secondly, two different people, or even the same person on two different days, may index similar images inconsistently. The result of these inefficiencies is a less than optimal search result for the end user of the system.

Computer do the indexing based on a CBIR scheme attempts to address the shortcomings of human-based indexing. Since a computer can process images at a much higher rate, while never tiring For example, each CBIR system needs to be tuned for its particular use in order to give optimal results. A retrieval system designed for querying medical x-ray images will more than likely prove to be a poor system for retrieving satellite images of South American rain forests. In addition, presently employed algorithms cannot yet consistently extract abstract features of images, such as emotional response, that would be relatively easy for a human to observe. Several approaches have been developed to capture the information of image contents by directly computing the image features from an image. The image features are directly constructed from the typical Block Truncation Coding or half toning based compressed data stream without performing the decoding procedure. These image retrieval schemes involve two phases, indexing and searching, to retrieve a set of similar images from the database. The indexing phase extracts the image features from all of the images in the database which is later stored in database as feature vector. In the searching phase, the retrieval system derives the image features from an image submitted by a user.

II RELATED WORK

Y. Gong and S. Lazebnik proposed the problem of learning binary codes that preserves the similarity for an efficient search for similarity in large-scale image collections is formulated by terms of zero-rotation data centering to minimizing quantization error by mapping data to the vertices of a zero-center binary hypercube as well as proposing a simple and efficient alternative minimizing algorithm to perform this operation [1].

The author Y. Pan, T. Yao, T. Mei, H. Li, C.-W. Ngo, and Y. Rui, proposed an approach for jointly exploring cross-view learning and the use of click data. The cross view learning is used for creating latent subspace with the ability to compare information from incomparable original views (ie text and image views), and use of click data explores access data that is widely available and freely accessible for understanding of the query [2].

The author D. Zhai, H. Chang, Y. Zhen, X. Liu, X. Chen, and W. Gao have been proposed HFL for the searching of inter-vision similarities. A new multimode HFL method, called Parametric Local Multimodal Hashing (PLMH) that can learn a set of hash functions to adapt locally to the data structure of each mode [3].

author G. Ding, Y. Guo, and J. Zhou proposed the problem of learning hash functions in the context of multimodal data for the search for similarity between cross-views is formulated by they proposed the Collective Matrix Factorization Hashing (CMFH) method which can generates unique hash codes for various modalities of single instance through collective matrix factorization along with the latent factor model [4].

Author H. Jegou, F. Perronnin, M. Douze overcommed the problem of large-scale image search. For this purpose they have provided three restrictions i.e search accuracy, effciency and memory usage and proposed different ways to add local image descriptors into a vector and demonstrated that Fisher's kernel performs as much better as visual bag approach for any given vector dimension [5].

The author J. Zhou, G. Ding, and Y. Guo proposed a new LSSH (Latent Semantic Sparse Hashing) algorithm to perform a search for similarity between modes using Sparse Coding and Matrix Factorization. For this purpose LSSH uses Sparse Coding to acquire the most important image structures and Matrix Factorization to learn the latent concepts of the text. [6].

The author Z. Yu, F. Wu, Y. Yang, Q. Tian, J. Luo, and Y.Zhuan proposed a Discriminative Coupled Dictionary Hashing (DCDH), in which the paired dictionary for each mode is acquired with secondary information (for example, categories). These coupled dictionaries not only preserve the intra-similarity and interconnection between multimode data, but also contain dictionary atoms that are semantically discriminating (that is, data in the same category are reconstructed from atoms in the similar dictionary) [7].

The author H. Zhang, J. Yuan, X. Gao, and Z. Chen has been proposed a method of cross-media recovery based on short and long-term relevance feedback. This method focused on two typical types of multimedia data, i.e. image and audio. Firstly they have created a multimodal representation through a statistical correlation between the image arrays and audio entities, and they defined the metric of the distance between the means for the measurement of similarity; therefore an optimization strategy based on relevant feedback combines the results of short-term learning and long-term accumulated knowledge in the objective function [8].

The author A. Karpathy and L. Fei-Fei proposed a model generating the descriptions of natural language of images and their regions. This approach have advantage of image data sets and their sentence descriptions to know the intermodal correspondences between language and visual data.The alignment model is based on combination of convolutional neural networks on image regions, bidirectional recurrent neural networks on sentences . The structured goal aligns two modalities through a multimodal model [9].

The author J. Song, Y. Yang, Y. Yang, Z. Huang, and H. T. Shen proposed a multimedia recovery paradigm to innovate large-scale research of different multimedia data. It is able to find results from different types of media of heterogeneous data sources, for example by using a query image to retrieve relevant text documents or images from different data sources [10].

III PROPOSED APPROACHES:-

We propose a secure framework for the storage and recovery of the subcontracted privacy protection in large archives of shared images. Our proposal is based on CBIR, a novel Encryption scheme of the image that presents image recovery properties based on content. The framework allows both encrypted storage and search using content-based image retrieval queries while preserving privacy against honest but curious cloud administrators. We have built a prototype of the proposed framework, formally analyzed and tested its safety properties, and experimentally assessed its performance and accuracy of recovery. Our results show that CBIR is probably safe, allowing more efficient operations that the existing proposals, both in terms of complexity of time and space, and opens the way to new scenarios of practical application.

Algorithms:

1. **AES Encryption Algorithm**

It is symmetric algorithm. It used to convert plain text into cipher text .The need for coming with this algo is weakness in DES/RSA. The 56 bit key of des is no longer safe against attacks based on exhaustive key searches and 64-bit block also consider asweak.Blowfish was to be used 128-bit block with 128-bit keys.

Input:

128_bit /192 bit/256 bit input (0, 1)

Secret key (128_bit) +plain text (128_bit).

Process:

10/12/14-rounds for-128_bit /192 bit/256 bit input

X or state block (i/p)

Final round: 10, 12, 14

Each round consists: sub byte, shift byte, mix columns, add round key.

Output:

Cipher text (128 bit)

2. Shape-Based Invariant Texture Index (SITI) for Feature Extraction

Feature Extraction image content identification.

Steps:

1. Color feature is one of the most widely used visual features in image retrieval, for its invariance with respect to image scaling, rotation, translation. In this work, an image is divided into four equal sized blocks and a centralized image with equal-size. For each block, a 9-D color moment is computed, thus the dimension of color comment for each image is 45. The 9-D color moment of an image segment is utilized, which contains values of mean, standard deviation and skewness of each channel in HSV color space.

2. Edge Detection: Most of the shape information of an image is enclosed in edges. So first we detect these edges in an image and by using these filters and then by enhancing those areas of image which contains edges, sharpness of the image will increase and image will become clearer.

Canny Edge Detection:

Canny edge detection is a technique to extract useful structural information from different vision objects and dramatically reduce the amount of data to be processed. It has been widely applied in various computer vision systems. Canny has found that the requirements for the application of edge detection on diverse vision systems are relatively similar. Thus, an edge detection solution to address these requirements can be implemented in a wide range of situations. The general criteria for edge detection include:

1. Detection of edge with low error rate, which means that the detection should accurately catch as many edges shown in the image as possible
2. The edge point detected from the operator should accurately localize on the center of the edge.
3. A given edge in the image should only be marked once, and where possible, image noise should not create false edges.

The Process of Canny edge detection algorithm can be broken down to 5 different steps:

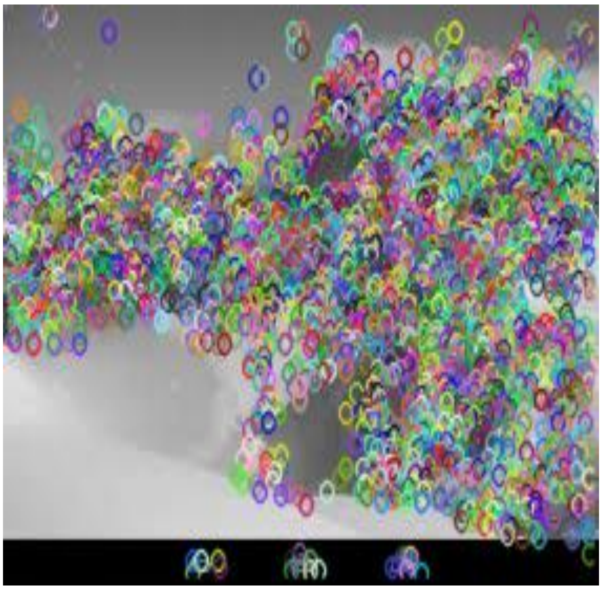
1. Apply filter to smooth the image in order to remove the noise
2. Find the intensity gradients of the image
3. Apply non-maximum suppression to get rid of spurious response to edge detection
4. Apply double threshold to determine potential edges
5. Track edge by hysteresis: Finalize the detection of edges by suppressing all the other edges that are weak and not connected to strong edges.

3. Texture feature describes the structure arrangement of surfaces and their relationship to the environment, such as fruit skin, clouds, trees, and fabric. The texture feature in our method is described by hierarchical wavelet packet descriptor (HWVP). A 170- D HWVP descriptor is utilized by setting the decomposition level to be 3 and the wavelet packet basis to be DB2.

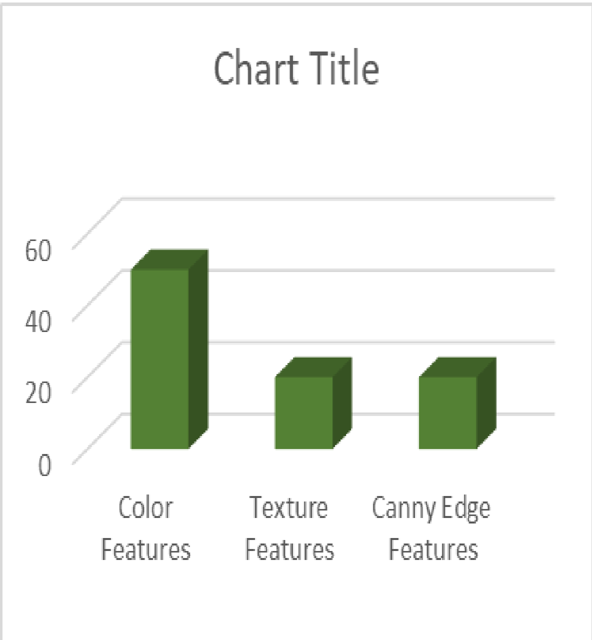
Working:

Input:

Output:



Feature Extraction:



Features	Percentage
Color	50 to 55
Texture	22 to 25
Canny Edge	15 to 18

IV RESULT AND DISCUSSION:

The experimental result evaluation, we have notation as follows:

TP: True positive (correctly predicted number of instance)

FP: False positive (incorrectly predicted number of instance),

TN: True negative (correctly predicted the number of instances as not required)

FN false negative (incorrectly predicted the number of instances as not required),

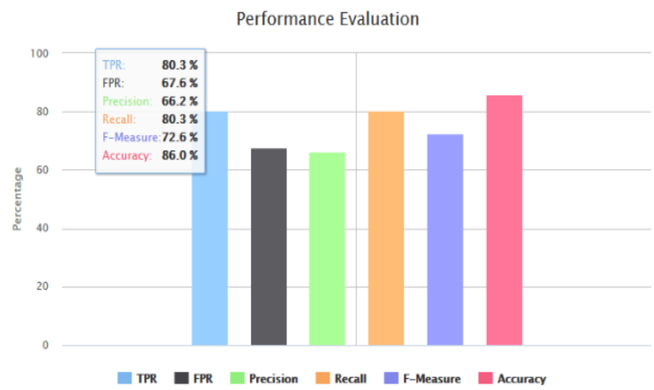
On the basis of this parameter, we can calculate four measurements

Accuracy = TP+TN÷TP+FP+TN+FN

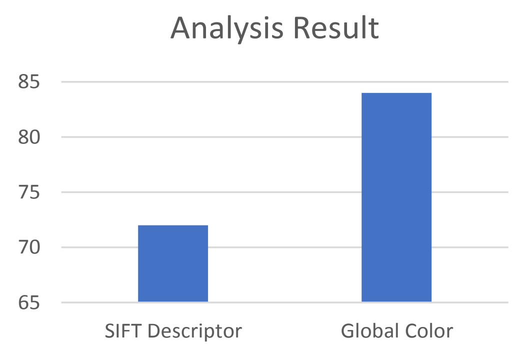
Precision = TP ÷TP+FP

Recall= TP÷TP+FN

F1-Measure = 2×Precision×Recall ÷Precision+ Recall



Parameters	Percentage
TPR	80.3%
FPR	67.6%
Precision	66.2%
Recall	80.3%
F-Measure	72.6%
Accuracy	86.0%



Comparison table

	IES-CBIR	SSE
Accuracy Percentage	84 to 87%	72%

V CONCLUSION

In this Paper, we have proposed a new secure framework for the external storage of privacy protection, research and recovery of large-scale dynamic image archives, where the

reduction of the general expenses of the customer is central appearance. At the base of our framework there is a new cryptography scheme, specifically designed for images, called CBIR. The key to its design is the observation that in the images, color information can be separated from the plot information, allowing the use of different cryptographic techniques with different properties for each and allowing to preserve privacy Image recovery based on the content that will be created from unreliable third-party cloud servers. We formally analyze the safety of our proposals and further experiments the evaluation of the implemented prototypes revealed that our approach reaches an interesting exchange between precision and I remember in the CBIR, while exhibiting high performances and scalability compared to alternative solutions.

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