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# **Efficient and Secure Image Reranking for TBIR**

G. Kavya<sup>1</sup>, P. Yaswanth<sup>2</sup>, P. Sathvik<sup>3</sup> <sup>1, 2, 3</sup> Department of IT, LBRCE

Abstract: Tag based image search plays an vital role to find images shared by clients in social networks. However, how to make the top ranked result relevant and with diversity. In previous papers these systems search image by using tags as query. However search results are weakly relevant tags and noisy tags. In this work we are using secure and efficient image re-ranking for TBIR. It overcomes the drawbacks that are present in the image re-ranking based on topic diversity.1) Tag mismatch 2) Query Ambiguity The main advantage of AES algorithm is to protect image data from unauthorized users. To detect and find unauthorized users for tag based image retrieval. AES is a type of symmetric key block cipher based on 10 rounds and key length is 128 bits .during encryption each round performs four transformations .sub bytes, shift rows, mix columns and add round key. AES is essential to secure the image data.

Keywords: AES algorithm, Image re ranking, tag based image retrieval, social media

## I. INTRODUCTION

In general there are 2 types of image retrieval process are present in the project, they are

#### A. Tag Based Image Retrieval system:

Tag-based image retrieval framework to improve the retrieval performance of a group of related personal images captured by the same user within a short period of an event by leveraging millions of training web images and their associated rich textual descriptions. For any given query tag the inverted file method is employed to automatically determine the relevant training web images that are associated with the query tag and the irrelevant training web images that are not associated with the query tag.

## B. Orientation

Orientation of edge pixels implies their direction. For instance, an edge may have a vertical orientation (pixels on a vertical line/curve) or a horizontal orientation (pixels on a horizontal line), or it may be slanted. You can represent the edge orientation by an angle .based on orientation it will display different images related to given query.

#### C. Working



Fig 1.1.1 semantic and visual information of images



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Fig 1.1.2.Tag based image retrieval

## D. Content Based Image Retrieval System

Content-based" means that the search analyzes the contents of the image rather than the metadata such as keywords, tags, or descriptions associated with the image. The term "content" in this context might refer to colours, shapes, textures, or any other information that can be derived from the image itself.

There is an explosion of social media content available online, such as Flicker, YouTube and Zoom. Such media repositories promote users to collaboratively create, evaluate and distribute media information. They also allow users to annotate their uploaded media data with descriptive keywords called tags. These valuable metadata can greatly facilitate the organization and search of the social media. By indexing the images with associated tags, images can be easily retrieved for given query. However, since user-provided tags are usually noisy and incomplete, simply applying text-based retrieval approach may lead to unsatisfactory results. Therefore, a ranking approach that is able to explore both the tags and images' content is desired to provide users better social image search results. Currently, Flicker provides two ranking options for tag-based image search. One is "most recent", which orders images based on their uploading time, and the other is "most interesting", which ranks the images by "interestingness", a measure that integrates the information of click-through, comments, etc. In the following discussion, we name these two methods time-based ranking and interestingness-based ranking, respectively. They both rank images according to measures (interestingness or time) that are not related to relevance and it results in many irrelevant images in the top search results. In addition to relevance, lack of diversity is also a problem. Many images from social media websites are actually close to each other. For example, several users get used to upload continuously captured images in batch, and many of them are visually and semantically close.

## II. RELATED WORK

In this paper [1] Social media sharing web sites like Flicker, Google etc allow users to comment on images with free tags, which is sufficient to facilitate Web image search an organization. However, the tags associated with an image generally are in a random order without any importance relevant information, which limits the efficacy of these tags in search and other applications and proposed a tag ranking scheme, aiming to automatically rank the tags associated with a given image according to their concernment to the image content. In this paper [2] Using Diverse Ranking scheme algorithm estimates the relevance scores of images with respect to the user query. By using Tag-based image search it frequently return results that are irrelevant or not diverse images. And they also used the greedy algorithm but it also return the irrelevant images. In this paper [3] Using the Multi Model Redundant Algorithm (MMR) System will return many non-relevant documents, and only very few (or none) relevant ones. To reduce this decompositions they used the other algorithm i.e. Relevant Novelty algorithm it returns the relevant ones. In this paper [4] By using previous MR approaches it doesn't retrieve accurate results. So, they used the other ranking algorithm i.e. two-step similarity ranking it displays the accurate results which are related to semantics and visual features of particular image. In this paper [5] Retrieve images are erroneous, incomplete and inconsistent. Using Social assisted media tagging (SMAT) it results complete and consistent data.



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#### III. EXISTING SYSTEM

In previous system using Tag based image search to retrieve low level features of the image. It does not support high relevance and diversity performance of initial retrieval results and ranking problem in the tag based image retrieval. tag based image retrieval achieved two challenges they are query ambiguity and tag mismatch. Social tagging requires users to tag their uploaded images with their own keywords and distribute with others. Different from ontology based image explanation; there is no predefined ontology in social image tagging. Every user has its own habit to tag images and retrieve irrelevant images based on query. So it will not provide security to retrieve images.

#### IV. PROPOSED WORK

Our proposed work Force which is a technique image Re-ranking using AES algorithm for TBIR. The main advantage of AES algorithm is to protect image data from unauthorized users. To detect and find unauthorized users for tag based image retrieval. AES is a type of symmetric key block cipher based on 10 rounds and key length is 128 bits .during encryption each round performs four transformations.1) sub bytes 2)shift rows 3)mix columns 4) add round key. AES is essential to secure the image data. Unlike DES, AES does not use Fiestel network, it is based on Rijndael. Rijndael has a key size of 128, 192 or 256 bits and a fixed block size of 128 bits which is larger than the DES block size. It provides the combination of permutation and substitution both. AES implementation can be done in both hardware and software. It provides the better security as compared to the DES, TDES, IDEA, BLOWFISH, etc. algorithms. It is free of cost. The algorithm provides faster encryption and flexibility as it implemented in both hardware.

## V. IMPLEMENTATION

In this system, in order to get the fruitful results we use AES algorithm and the modules involves in this system are stated below

#### A. AES Algorithm

The advance encryption standard (AES) specifies federal information processing standards publication (FIPS) approved cryptographic algorithms that can be used to protect electronic data. To detect and find unauthorized users for tag based image retrieval. AES algorithm is of three types i.e. AES-128, AES-192 and AES-256. This classification is done on the bases of the key used in the algorithm for encryption and decryption process. The numbers represent the size of key in bits. This key size determines the security level as the size of key increases the level of security increases. The AES algorithm uses a round function that is composed of four different byte-oriented transformations. AES is a type of symmetric key block cipher based on 10 rounds and key length is 128 bits .during encryption each round performs four transformations. AES is essential to secure the image data.

B. Steps

To input an Image. Convert this analog image into digital image. Convert the image pixels into Row and Columns form. Perform transpose of Columns. Add these columns using bitwise-Xor operation. Get the Output.

- C. Steps involved in encryption process:
- 1) Substitute byte
- 2) Shift ro
- 3) Mix column
- 4) Add round key

Decryption process is the reverse of encryption process.

D. Steps involved in decryption process:

- *1)* Inverse shift row
- 2) Inverse substitute byt
- 3) Add round k



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## *4)* Inverse mix columns

There is a number of round present of key and block in the algorithm. The number of rounds depends on the length of key use for encryption and decryption.

	TABLE 1		
Key-Block Round combinations			
	Key length(in word/byte/bits)	Block size(in	Numbers of rounds
		word/byte/bits)	
AES-128	4/16/128	4/16/128	10
AES-192	6/24/192	4/16/128	12
AES-256	8/32/256	4/16/128	14



5.1.1. Flowchart for Encryption process



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5.1.2. Flowchart for Decryption process

## E. Methodology



First the user asks a query image to the database that he wants to retrieve relevant images. He gives the input as query then the database starts the query processing by taking the input as text or number query. Now database searches related images that are related to input query. In this first it constructs graph by using tag graph construction. Tag graph construction based on the tag information of image dataset. Tag graph is constructed to mine the topic community. Community detection. Affinity propagation clustering method is employed to detect topic communities. Image community mapping process. We assign each image to a single community according to the tag overlap ratio between the topic community and image. Inter-community ranking. We introduce the adaptive random walk model to rank topic communities according to the semantic relevance between the community and query. Intra-community ranking. A regularization framework is proposed to determine the relevance of each image to the query by fusing the visual, semantic and view information into a unified system. Finally it will display related images which are related to user query but not accurate.



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5.1.3. Flow chart for AES algorithm

## VI. EXPERIMENTAL results

Tag based image search plays an vital role to find images shared by clients in social networks. using this technique identify related images based on given query but not accurate. retrieval images access by unauthorized user and not provide the security. using AES algorithm is essential to protect the confidential image data from unauthorized users.



6.1 Represents Mat lab window



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6.2. Represents query image



6.3. Represents orientation of image



6.4. Tag graph construction process completed





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6.5. Affinity clustering for community detection



#### 6.6. Represents related images



6.6. AES algorithm to provide security



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#### VII. CONCLUSION

In this paper, Image Encryption and Decryption using AES algorithm is implemented to secure the image data from an unauthorized access. A Successful implementation of symmetric key AES algorithm is one of the best encryption and decryption standard.

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