# Review on Content Based Image Retrieval (CBIR) Technique

Prof. Vikram M Kakade<sup>1</sup>, Ishwar A. Keche<sup>2</sup>

Lecturer, Department of Electronics & Telecommunication Engineering, PRMCEAM,-Badnera, Amravati UG Student, Department of Electronics & Telecommunication Engineering, PRMCEAM,-Badnera, Amravati

Abstract: In digital image processing finding or searching a image in huge database of digital images is big task for designing more efficient image and shape matching algorithm. In data storage and image acquisition technologies we have to enabled the creation of large image datasets. In this type, it is necessary to develop appropriate information systems to efficiently manage these collections. The common approaches use the so-called Content-Based Image Retrieval (CBIR) systems. Basically, these systems try to retrieve images similar to a user-defined specification or pattern (e.g., shape sketch, image example). In this paper we see the different type's techniques of content based image retrieval for different application purpose

Keywords: content-based image retrieval, image database, image descriptors, indexing, query specification, query visualization, effectiveness measures, color, texture, shape, Semantic retrieval, Relevance feedback

## **1** Introduction

Content-based image retrieval (CBIR), also known as query by image content (QBIC) and content-based visual information retrieval (CBVIR) is the application of computer vision techniques to the image retrieval problem, that is, the problem of searching for digital images in large databases (see this survey for a recent scientific overview of the CBIR field). Content-based image retrieval is opposed to traditional concept-based approaches (see Concept-based image indexing)." 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 colors, shapes, textures, or any other information that can be derived from the image itself. CBIR is desirable because searches that rely purely on metadata are dependent on annotation quality and completeness. Having humans manually annotate images by entering keywords or metadata in a large database can be time consuming and may not capture the keywords desired to describe the image. The evaluation of the effectiveness of keyword image search is subjective and has not been well-defined. In the same regard, CBIR systems have similar challenges in defining success

## 2 Overview

The term "content-based image retrieval" seems to have originated in 1992 when it was used by T. Kato to describe experiments into automatic retrieval of images from a database, based on the colors and shapes present. Since then, the term has been used to describe the process of retrieving desired images from a large collection on the basis of syntactical image features. The techniques, tools, and algorithms that are used originate from fields such as statistics, pattern recognition, signal processing, and computer vision. The earliest commercial CBIR system was developed by IBM and was called QBIC (Query by Image Content). Recent network and graph based approaches have presented a simple and attractive alternative to existing method. The interest in CBIR has grown because of the limitations inherent in metadata-based systems, as well as the large range of possible uses for efficient image retrieval. Textual information about images can be easily searched using existing technology, but this requires humans to manually describe each image in the database. This can be impractical for very large databases or for images that are generated automatically, e.g. those from surveillance cameras. It is also possible to miss images that use different synonyms in their descriptions. Systems based on categorizing images in semantic classes like "cat" as a subclass of "animal" can avoid the mix categorization problem, but will require more effort by a user to find images that might be "cats", but are only classified as an "animal". Many standards have been developed to categorize images, but all still face scaling and mix categorization issues. Initial CBIR systems were developed to search databases based on image color, texture, and shape properties. After these systems were developed, the need for user-friendly interfaces became apparent. Therefore, efforts in the CBIR field started to include human-centered design that tried to meet the needs of the user performing the search. This typically means inclusion of: query methods that may allow descriptive semantics, queries that may involve user feedback, systems that may include machine learning, and systems that may understand user satisfaction level



Figure 1: General scheme of content-based image retrieval

## **3** Techniques of CBIR

Many CBIR systems have been developed, but the problem of retrieving images on the basis of their pixel content remains unsolved

#### 1) Query techniques or Query based techniques

Different implementations of CBIR make use of different types of user queries.

Query by example is a query technique that involves providing the CBIR system with an example image that it will then base its search upon. The underlying search algorithms may vary depending on the application, but result images should all share common elements with the provided example See also called as Reverse image search

Options for providing example images to the system include

- A preexisting image may be supplied by the user or chosen from a random set.
- The user draws a rough approximation of the image they are looking for, for example with blobs of color or general shape
- *a)* Semantic retrieval

Semantic retrieval starts with a user making a request like "find pictures of Abraham Lincoln". This type of open-ended task is very difficult for computers to perform - Lincoln may not always be facing the camera or in the same pose. Many CBIR systems therefore generally make use of lower-level features like texture, color, and shape. These features are either used in combination with interfaces that allow easier input of the criteria or with databases that have already been trained to match features (such as faces, fingerprints, or shape matching). However, in general, image retrieval requires human feedback in order to identify higher-level concepts.

b) Relevance feedback (human interaction)

Combining CBIR search techniques available with the wide range of potential users and their intent can be a difficult task. An aspect of making CBIR successful relies entirely on the ability to understand the user intent. CBIR systems can make use of relevance feedback, where the user progressively refines the search results by marking images in the results as "relevant", "not relevant", or "neutral" to the search query, then repeating the search with the new information. Examples of this type of interface have been developed.

c) Iterative/machine learning

Machine learning and application of iterative techniques are becoming more common in CBIR

d) Other query methods

Other query methods include browsing for example images, navigating customized/hierarchical categories, querying by image region (rather than the entire image), querying by multiple example images, querying by visual sketch, querying by direct specification of image features, and multimodal queries (e.g. combining touch, voice, etc.)

#### 2) Content comparison using image distance measures

The most common method for comparing two images in content-based image retrieval (typically an example image and an image from the database) is using an image distance measure. An image distance measure compares the similarity of two images in various dimensions such as color, texture, shape, and others. For example, a distance of 0 signifies an exact match with the query, with respect to the dimensions that were considered. As one may intuitively gather, a value greater than 0 indicates various degrees of similarities between the images. Search results then can be sorted based on their distance to the queried image. Many measures of image distance (Similarity Models) have been developed.

#### a) Color

Computing distance measures based on color similarity is achieved by computing a <u>color histogram</u> for each image that identifies the proportion of pixels within an image holding specific values. Examining images based on the colors they contain is one of the most widely used techniques because it can be completed without regard to image size or orientation. However, research has also attempted to segment color proportion by region and by spatial relationship among several color regions.

b) Texture

Texture measures look for visual patterns in images and how they are spatially defined. Textures are represented by texts which are then placed into a number of sets, depending on how many textures are detected in the image. These sets not only define the texture, but also where in the image the texture is located.

Texture is a difficult concept to represent. The identification of specific textures in an image is achieved primarily by modeling texture as a two-dimensional gray level variation. The relative brightness of pairs of pixels is computed such that degree of contrast, regularity, coarseness and directionality may be estimated. The problem is in identifying patterns of co-pixel variation and associating them with particular classes of textures such as silky, or rough.

Other methods of classifying textures include:

- Co-occurrence matrix
- Laws texture energy
- Wavelet transform
- Orthogonal transforms (Discrete moments)
- c) Shape

Shape does not refer to the shape of an image but to the shape of a particular region that is being sought out. Shapes will often be determined first applying segmentation or edge detection to an image. Other methods use shape filters to identify given shapes of an image. Shape descriptors may also need to be invariant to translation, rotation, and scale.

### 4 Applications

Potential uses for CBIR

- Architectural and engineering design
- Art collections
- Crime prevention
- Geographical information and remote sensing systems
- Intellectual property
- Medical diagnosis
- Military
- Photograph archives

# 5 Conclusion

In this paper we have reviewed complete CBIR Technique on QBIC for which shape, color & Texture are taken in analysis with different retrieval technique used for digital image matching in Digital image processing.

# References

- Content-based Multimedia Information Retrieval: State of the Art and Challenges (Original source, 404'd)Content-based Multimedia Information Retrieval: State of the Art and Challenges, Michael Lew, et al., ACM Transactions on Multimedia Computing, Communications, and Applications, pp. 1–19, 2006.
- 2. Eakins, John; Graham, Margaret. "Content-based Image Retrieval". University of Northumbria at Newcastle. Retrieved 2014-03-10.
- Rui, Yong; Huang, Thomas S.; Chang, Shih-Fu (1999). "Image Retrieval: Current Techniques, Promising Directions, and Open Issues". Journal of Visual Communication and Image Representation. 10: 39– 62. doi:10.1006/jvci.1999.0413. Retrieved 2 July 2016.
- 4. Shapiro, Linda; George Stockman (2001). Computer Vision. Upper Saddle River, NJ: Prentice Hall. ISBN 0-13-030796-3.
- Datta, Ritendra; Dhiraj Joshi; Jia Li; James Z. Wang (2008). "Image Retrieval: Ideas, Influences, and Trends of the New Age". ACM Computing Surveys. 40 (2): 1– 60. doi:10.1145/1348246.1348248.

# **Author Profile**

<b>Prof.Vikram M Kakade</b> Had completed his PG Degree in Electronics & Telecommunication Engineering & Working as a Lecturer in EXTC Department of PRMCEAM from Last 5 years .His area of working includes Digital Image Processing, Embedded System & Control System & Automation. Role of Him in this project as a Guide on UG Final Project of 2 <sup>nd</sup> author.
<b>Ishwar A. Keche</b> Is Student of Final Year Electronics & Telecommunication Engineering of PRMCEAM Working on Image Matching project under guidance of 1 <sup>st</sup> author his area of interest includes Algorithms of Image Matching.