

## Late Semantic Fusion Approaches For Multimedia Information Retrieval

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**Abstract:** In retrieval, indexing and classification of multimedia information, an efficient the features for image retrieval are quantitatively compared and their correlation is analyzed. We focused on applying different strategies of merging multimedia information, textual and visual, following the late fusion approaches. There are number of fusion approaches are used to have different modalities are used to accomplish various multimedia related analysis tasks. The problem of multimedia based image retrieval search system a very rapid growth in the quantity and availability of the digital images involves the researches into the multimedia information system. Main aim of this work is to improve multimedia information retrieval task by using textual pre-filtering combined with an image re-ranking. We will try to improve the result to facilitate generating high quality social tags; tag recommendation by automatically assigning relevant tag to a photo draws a particular research interest.

**Keyword-** Multimedia information fusion, multimedia retrieval, tag recommendation, late fusion, image re-ranking.

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### I. INTRODUCTION

With the advances in the different computer technologies and the development and advent in the world wide web, there has been an explosion in the amount and the complexity of the digital data being generated, stored, transmitted, analyzed and accessed[1]. The Multimedia based Information Retrieval System (MIRS) is the system which is able to store, retrieve, maintain, update and manage the information in the database. In Multimedia information retrieval system, there is difficulty in the communication between an information or images, users and the image retrieval system.

The users may contain the different needs and knowledge about the images collection in database and an image retrieval system, it must support different forms of query formulation in the system. When an image is observed there can be problem of semantic gap which defines “gap between the image in the real world and the information consist in a description which is resulting from a recording of that different scene for different image”. That is, when an image is provided to different dataset, from the image some information was present in the real world is automatically missing. So, the loss of information can be due to the missed details, bad clarification or different viewing angles or any type of imperfectness of an image capturing device like a camera.

In this paper we discussed, Semantic filtering method that seeks to enhance the similarities between multimedia items when they are composed of both a visual and a textual part. We discuss how such a filter based on the text query can better cope with the semantic gap in the case of CBMIR. We propose to

use this approach as a first level of the fusion process of visual and textual information in our multimedia relevance model.

## **II. LITERATURE REVIEW**

Multimedia information retrieval of the different text, images, audio, and video are may be the best developed technology. As early in 1986, image databases were being developed and deployed the system such as UC Berkeley's Image Query on system, so "Developers believe that this software was the first deployed multi-user networked digital image database system in the environment".[3] In the recent times, multimodal fusion techniques have gained much attention of many researchers due to the benefit it provides for various multimedia analysis tasks. One of the earliest considerations is to decide what strategy to follow when fusing multiple modalities. The most widely used strategy is to fuse the information at the feature level, which is also known as early fusion.[4][5] The other approach is decision level fusion or late fusion which fuses multiple modalities in the semantic space. A combination of these approaches is also practiced as the hybrid fusion approach.[6]

Multimedia Information Retrieval is usually addressed from a textual point of view in most of the existing commercial tools, using annotations or metadata information associated with images or videos. In this work we deal with both textual and visual information, carrying out both monomodal and multimodal experiments using different multimedia fusion techniques and algorithms. Multimedia fusion tries to use the different media sources as complementary information to increase the accuracy of the retrieved results, in order to help in solving the semantic gap problem, referred to the difficulty in understanding the information that the user perceives from the low level characteristics of the multimedia data. Specifically, in the case of Image Retrieval, the semantic gap is the lack of correspondence between the information from visual features (e.g., histograms) and the interpretation of these data by a user in a certain situation (visually similar images to the query in terms of low level features can be very different in terms of meaning).The benefits of multimedia fusion come from approaches that improve the results of the monomodal search, balancing the cost and complexity of the implementation and deployment and providing correct and complementary information to the monomodal results. When multimedia approaches are used, several aspects have to be taken into account in order to select the most appropriate. "The CLEF 2011 Photo Annotation and Concept-based Retrieval Tasks" The experiments have been carried out using the Wikipedia collection at ImageCLEF 2011 that contains almost 240 thousand socially annotated images in their Wikipedia articles (also provided), 50 multimedia topics and the related relevance judgments. Both task differentiate among approaches that consider solely visual information, approaches that rely only on textual information in form of image metadata and user tags, and multimodal approaches that combine both information sources. The relevance assessments were acquired with a crowdsourcing approach and the evaluation followed two evaluation paradigms: per concept and per example. Results indicate that while the annotation task shows promising results, the concept-based retrieval task is much harder to solve, especially for specific information needs. A closer examination showed that all approaches had great difficulties to identify relevant images for the topic female old person.

## **III. PROPOSED SYSTEM**

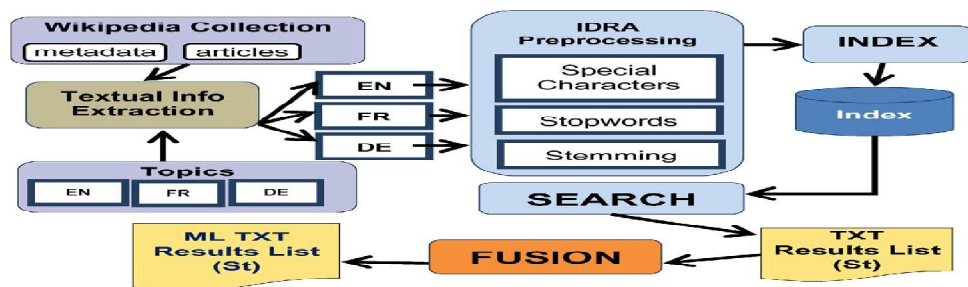
The TBIR subsystem acts firstly over the whole images of the database, acting as a filter to the CBIR system selecting the relevant images for a certain query.

In a second step, the CBIR system works over the set of filtered images reordering this list taking into account the visual information of the image.

The CBIR system generates different visual result lists depending on the number of query Images.

### A. Text-based Index and Retrieval

This module is in charge of the textual image retrieval using the metadata supplied for the images in the collection. IDRA tool extracts, selects, preprocesses and indexes the metadata information, for later search and retrieve the most relevant images for the queries. After this process, a ranked results list is obtained for each textual experiment.



“Figure 1. Text-based Index and Retrieval”

### B. Content Based Information Retrieval

Content-based image retrieval, a technique which uses visual contents to search images from large scale image databases according to users' interests, has been an active and fast advancing research area. In this case of retrieval system we need to focus on the metadata of the given objects. Here it requires having an index prepared like structure which will hold the metadata related with all the objects or images in case of our proposed system.

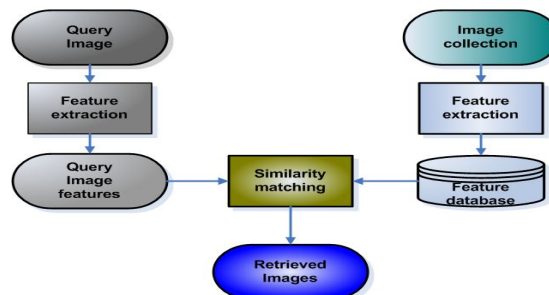


Figure 2. System Architecture

### C. Feature Extraction:

The visual low-level features for all the images in the database for the example images for each topic are extracted using the CEDD given by the Image CLEF 2011 organization. The CEDD descriptors, which include more than one feature in a compact histogram (color and texture information), belong to the family of Compact Composite Descriptors. The structure of CEDD consists of 6 texture areas. In particular, each texture area is separated into 24 sub-regions, with each sub-region describing a color.

Figure 2. Content based image retrievals

#### D. Similarity Module:

The similarity module uses our own logistic regression relevance feedback algorithm to calculate the Similarity ( $S_i$ ) of each of the images of the collection to the query. The algorithm calculates the probability of an image belonging to a set of those images sought by the query, and models the *logit* of this probability as the output of a generalized linear model whose inputs are the visual low-level image features.

### IV. MATHEMATICAL MODEL

$U = \{K, C, S, R, F\}$

Where  $K = \{S_1, S_2, S_3, \dots, S_n, S_n \neq 0\}$

where K is set of keyword.

Where  $C = \{P_1, P_2, P_3, \dots, P_n, P_n \neq 0\}$

where C is set of ranking function.

Where  $S = \{C_1, C_2, C_3, \dots, C_n\}$

where S is set of search.

Where  $R = \{M_1, M_2, M_3, \dots, M_n\}$

where R is set of result.

Where  $F = \{T_1, T_2, T_3, \dots, T_n\}$

where F is set of final result.

### V. RESULT AND ANALYSIS

As we are saying that we can retrieve images as soon as user will enter a keyword for searching any image. Therefore need arise to calculate the ranking of given keyword which is then use with respect to image searching technique.

Following snapshot depicts the starting module of our proposed system in which searching is provide on the basis of file name based path will be specified where all those images are to stored and then after Specifying same system will start further operation.

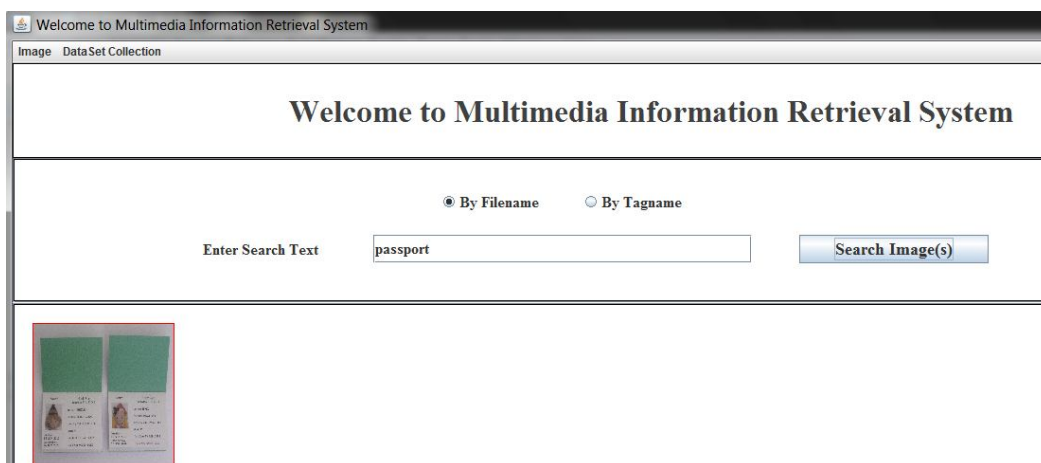


Figure 3. Search by file name.

In the next snapshot we are showing actually how system is operating on Tag based search method. As we are saying that system will analyze all the images present in the collection.



Figure 4 .Search by tag name

## CONCLUSION & FUTURE SCOPE

The proposed application is to develop an image retrieval application which can perform identity check of an image. The objective is to progress towards the user fulfillment by returning images that have a higher probability to be accepted, meaning is that those images will be highly relevant to the user's query. Each feature defines a multidimensional space where images are points, and the similarity between images is computed as the distance between points. Therefore we need to focus on various aspects related with the image i.e. we need to focus on content dependent and content independent data related with image.

The main goal of this work is to provide detailed description and analysis of textual pre-filtering techniques. These textual pre-filtering techniques reduce in a suitable way the size of the multimedia database improving the final fused retrieval results. Experiments show that the combination of textual pre-filtering and image re-ranking lists in a late fusion algorithm outperforms those without pre-filtering. It seems that textual information better captures the semantic meaning of a topic and that the image re-ranking fused with the textual score helps to overcome the semantic gap. With respect to the late fusion algorithms analyzed, better results are obtained with those that work only with the value scores than others, which rely on the ranked positions.

In the future, well develop new methods to speed the re-ranking processes in large-scale visual search systems. Beyond the visual features used in this work, well also explore the use of a large set of generic concept detectors in computing shot similarity or multimedia document context.

## APPENDIX

Paper Titled "A Survey on Multimedia Information Retrieval System" International Journal of Computer Technology and Application(IJCTA), Volume 5 . ISSN: 2229-6093, 1945-1949.

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