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# Enhancement of Text Based Image Retrieval to Expedite Image Ranking

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## ABSTRACT

*The monstrous development of advanced images over the web required the best image retrieval procedures that can enhance the recovery precision of the images. Hence research focus has been shifted from designing of sophisticated algorithms that can lessen the semantic gap between visual elements and extravagance of human semantics. Consequently numerous images reranking strategies has been proposed to improve the text based image results by exploiting visual data contained in the images. Semantic attributes are high level semantics properties are exploited by using the hyper graph re ranking method. A hypergraph model the relationship between the images as per the relevance score to order the images.. Attribute based image representation has achieved good performance in assisting the work of image classification with hypergraph. Hence refinement of Image search using attribute assisted re ranking method is addressed in this project*

## I. Introduction

With the substantial increase of online images, image retrieval has attracted significant attention in both academia and industry. The existing web image search engines including Google and Yahoo retrieve and rank images mostly based on textual information associated with the image in hosting web pages, such as the title and the surrounding text. In any case, text based image recovery experiences vital troubles that are brought about for the most part by the lack of ability of the related content to suitably portray the image content. As of late, visual reranking has been offered to refine text based list items by abusing the visual data contained in the images. To improve the accuracy of the text based image search ranking, visual re ranking has been put forth to refine the search results from text based image search engine by incorporating the information conveyed by the visual modality.

The current visual reranking strategies can be commonly sorted into three categories classification based, clustering based and graph based methods. In the classification based techniques, visual reranking is figured as binary classification to identify whether each search outcome is relevant or not. Pseudo Relevance Feedback (PRF) is applied to select set of training images to know the classifier. In numerous genuine situations, delegate cases acquired through PRF for preparing the dataset are exceptionally noisy and won't not be satisfactory for building up effective classifiers.

Graph based techniques have been proposed as of late and got increasing considerations. The multimedia entities in top ranks and their visual relationship can be spoken to as a gathering of nodes and edges. The salient features found using graph analysis is intense to enhance the viability of rank records. By the way, the reranking calculations said above are absolutely taking into account low-level visual elements, while for the most part don't consider any semantic relationship among initial ranked list.

## II. LITERATURE SURVEY

In this chapter, we provide a brief description of the various existing search reranking approaches exploited in the recent literature.

**“Bayesian video search reranking”*Transaction on Multimedia*X. Tian, L. Yang**

This related work simply shows the overall description about the existing working system. The existing system is working on the integrating the visual features and the attribute to the image searching. The system review in recent literature, and the quite knowledge about the hyper graph learning theory .It uses Low level feature extraction method The advantages of this is to improve the accuracy of visual reranking. The Disadvantages is searching methodology is not efficient.

**“Harvesting image databases from the web”** by **F. Jing and S. Baluja** develop Visual rank to analyze the visual link structures among images. The images found to be “authorities” are chosen as those that answer the image-queries well. To understand the performance in a real system, we conducted a series of large scale experiments on the basis of the task of retrieving images. It improve user satisfaction and relevancy result as compare to the result of Google Image Search Maintaining modest computational cost is vital to ensuring that this procedure can be used in practice; CBIR and Eigen Vector method is use The advantages of this is for quantifying the effectiveness of visual features by using bias vector visual rank is computed. It is not showing the relationship between the image similarity and likelihood for transaction more extensively is the disadvantages.

**“Image ranking and retrieval based on multi-attribute queries”** by **Farhadi, I. Endres**

The propose learning attributes which represents a new challenge is generalization across object categories is not just across instances within a category. It introduce a simple feature selection method for learning attributes which are generalize well across different categories. Evaluation that provides insights into the limitations of the standard recognition model of naming and introduce the new abilities provided by attribute based framework

### III. EXISTING SYSTEM

The existing text based image system is generally successful to hunt down significant images; the exactness of the query output is to a great extent limited by the bungle between the genuine pertinence of an image and its importance induced from the related textual descriptions. To enhance the rightness of the text based image search visual reranking has been advanced to refine the search result gotten from the text based image web crawler by consolidating the data bring by the visual methodology.

The current visual reranking techniques can be by and large sorted into three classifications as the clustering based, classification based and graph based strategies. In clustering analysis estimates the inter-entity similarity. The images are primarily grouped into many near duplicate media documents. However for queries that return highly diverse results or without clear patterns, the performance hence is not guaranteed.

In the classification based techniques, visual reranking is figured as binary classification to identify whether each search outcome is relevant or not. Pseudo Relevance Feedback (PRF) is applied to select set of training images to know the classifier.

Be that as it may, in numerous genuine situations, delegate cases acquired through PRF for the preparing dataset are exceptionally noisy and won't not be satisfactory for building up effective. Graph based techniques have been proposed as of late and got expanding consideration as exhibited to be viable. The multimedia entities in top ranks and their visual relationship can be spoken to as a gathering of nodes and edges. The salient features found using graph analysis is intense to enhance the viability of rank records. By the way, the reranking calculations said above are absolutely taking into account low-level visual elements while for the most part don't consider any semantic relationship among initial ranked list.

### IV. PROPOSED SYSTEM

The proposed method is to refine text-based search results by exploiting the visual information contained in the images. Attributes are expected to narrow down the semantic gap between low level visual and high level semantic meaning. Based on different classifier for all predefined attributes, each image is represented by an attribute feature. Attribute assisted hypergraph learning technique is utilized to reorder the positioned pictures which came back from web crawler in view of textual query. A hyper graph is reconstructed to model the

relationship of all the images, in which each vertex denoted an image and a hyperedge represents an attribute. A hyperedge in hypergraph is able to link more than two vertices. The weight of each edge is based on visual and attribute similarities of images belongs to the edge. The resultant set of images thus enhances the ranking performance.

### V. System Architecture

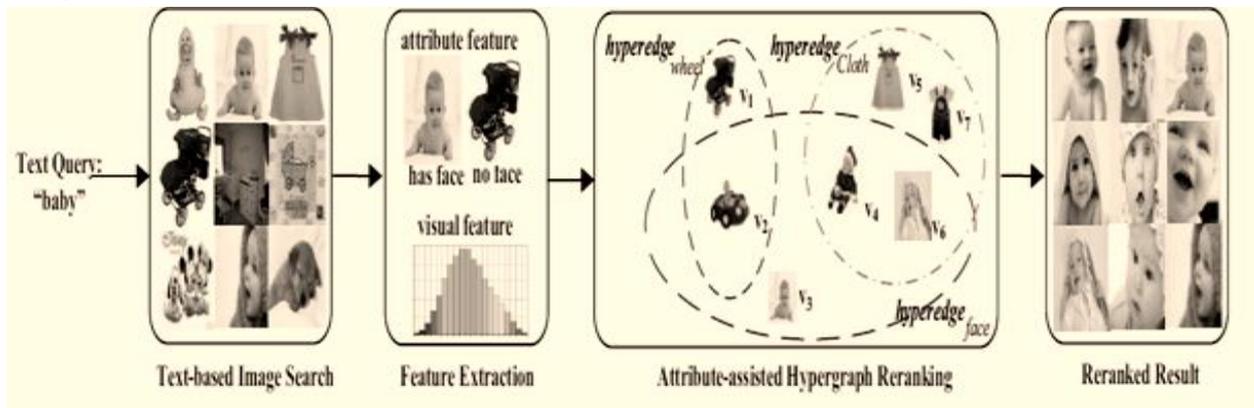


Figure 5.1 Architecture

The proposed system architecture is depicted in the figure 5.1, after a query “baby” is fired; an initial result is achieved via a text-based search engine. It is identified that text based search oftentimes yields inconsistent results. Some outwardly comparable to images are circulated in the results while other irrelevant results are filled between them, for example, "puppy" and "disney baby". In light of the returned images, both visual components and attribute elements are extracted. Specifically, the attribute feature of every image comprises of the reactions from the paired classifiers for all the characteristics. Visual representation and semantic properties are at the same time abused in a bound together model called hypergraph. Hypergraph is reproduced to demonstrate the relationship of all the images, in which each vertex denotes an image and a hyperedge represents an attribute and a hyperedge joins to several vertices. The weight of each edge is based on visual features and attributes similarities of images. The re ranked results gives the top nine best results of the system.

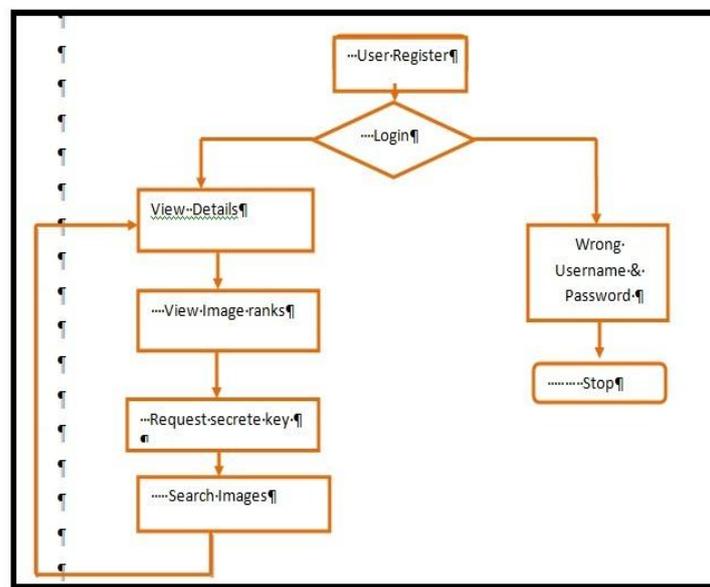
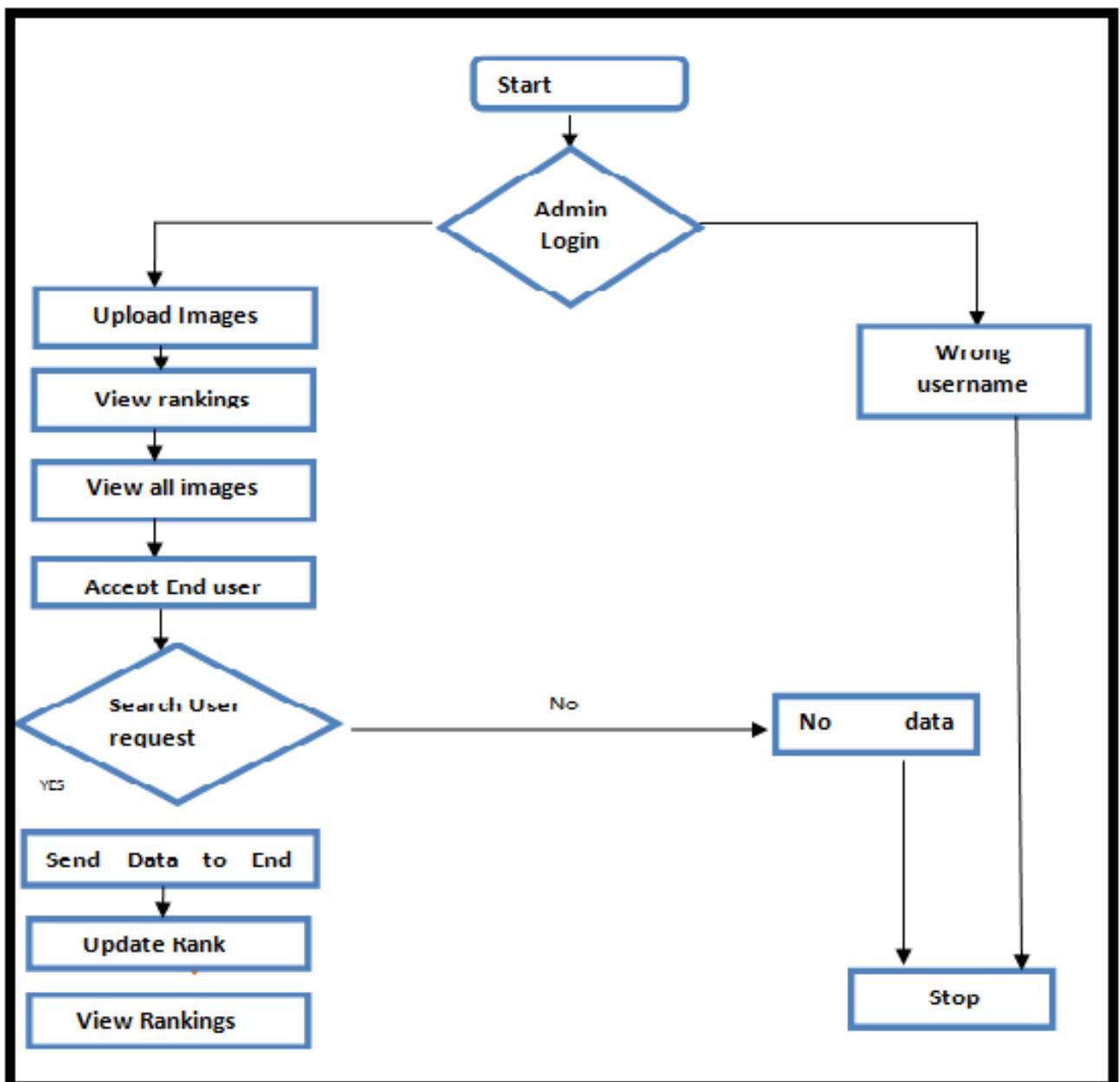


Fig 5.1 Process Flow Chart of USER

**Admin Process**

The figure 5.2 (b) depicts the flow diagram of the admin process. The Admin logs admin name and password, after login successful he can do some operations such as upload images, view uploaded images, view the searching history, view all image ranking and view all users, search images and logout. The admin can view the search history details. If he clicks on search history button, it will show the list of searched user details with their tags such as user name, user searched for image name, time and date. The admin can view the list of ranking images. If admin click on list of ranking images, then the server will give response with their tags image and rank of image. The admin can upload n number of images. Admin want to upload new image then he has enter some fields like image name, image color, image description, image type, image usage, browse the image file and upload.



**Figure 5.2 Process Flow chart of Admin**

## VI. RESULTS AND DISCUSSION

The experimental data used is MSRA-MM V2.0 dataset. This dataset consists of about 1 million real world image set and is publicly available. There are 900 images for each query and for each image its relevancy is labeled with three levels, very relevant, relevant and irrelevant. The Figure 8.1 illustrates images of “Steve Jobs”, “Cattle”, “Airplane” with different relevancy levels.



Figure 8.1 Image Analysis Examples

The Evaluation Measure adopted is in terms of Normalized Discounted Cumulative Gain (NDCG). NDCG is the standard evaluation in information access when they are two relevance levels, for measuring the performance.

With the ranking list given the NDCG score at position n is defined as

$$NDCG @n = z_n \sum_{j=1}^n \frac{2^{r(j)} - 1}{\log(1 + j)}$$

$Z_n$  is normalization factor,  $r(j)$  is the relevance score of  $j$ th image. Hence to evaluate overall performance, average of the NDGCs over all queries is obtained as Mean NDGC.

The performance in terms of MNDGC of ranking method with the text baselines are given in the table 8.1. We can observe that hypergraph reranking achieves around 3.8% improvement at MNDGC@20.

Table 8.1 Performance comparison for hypergraph reranking

MNDGC	20	40	60	80
Text baseline	0.45	0.43	0.42	0.40
Hypergraph	0.55	0.54	0.53	0.52

The Figure 8.2 illustrates the attributes that have top 5 highest weights



Face  
Head  
Nose  
Eye  
Ear



Wheel  
Vehicle part  
Shiny  
Headlight  
Metal



**Figure 8.2 Visualized search examples**

Case study of the visualized search examples between the proposed method and baselines of Bayesian reranking and cluster based approach as shown in Figure 8.3.

The first row gives depicts the proposed method, second row are results from Bayesian reranking and the third row is cluster based approach. As we can see in the second and third row, there are sets of irrelevant images in the result set. These results clearly depict that the proposed approach significantly outperforms the other existing methods.

## VII. CONCLUSION AND FUTURE WORK

### 7.1 Conclusion

Image search reranking has been calculated for several years and various approaches have been developed recently to boost the performance of text-based image search engine. This project serves as an attempt to include the attributes in reranking framework. We observe that semantic attributes are expected to narrow down the semantic gap between low-level visual features and high-level semantic meanings. We conclude that hypergraph reranking with the attributes assisted retrieval gives a significant performance over the other methods.

### 7.2 Future Work

An efficient combination by low-level visual features and semantic attribute features, done in attribute hypergraph learning approach can be further worked on the multi attributes and also a much better performance can be achieved.

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