

# Face Ascertain Based Method For Accessible By Pardons Words



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**ABSTRACT:** Now a days the popularity of social networks like face book, twitter are mostly used by the people. Many of them use human face images to their profile. And also we can maintain large scale database for the image storage. To avoid the large database use two algorithms like attribute enhanced sparse codewords and attribute embedded inverted indexing and used in offline and online storage respectively. In large image database have problem regarding the image retrieval .By using this algorithm we can efficiently retrieve the images from the large image database. It will give the 80% perfect matched images.

**Keywords:** Attribute enhanced sparse codewords, attribute embedded inverted indexing, ranking, attributes, annotation.

## I. INTRODUCTION

To analyze the effectiveness of different human attributes across datasets and find informative human attributes. To sum up, the contributions of attribute enhanced sparse codewords include: In the combine automatically detected high-level human attributes and low-level features to construct semantic codewords. To the best of developer knowledge, this is the first proposal of such combination for content-based face image retrieval. To balance global representations in image collections and locally embedded facial characteristics, in the scalable face image retrieval using attribute enhanced sparse codewords system two orthogonal methods to utilize automatically detected human attributes to improve content-based face image retrieval under a scalable framework. It conducts extensive experiments and demonstrates the performances of the proposed methods on two separate public datasets and still ensures real time response. In the further identify informative and generic human attributes for face image retrieval across different datasets. The selected descriptors are promising for other applications.

Automatically detected human attributes have been shown promising in different applications recently. Kumar et al. propose a learning framework to automatically find describable visual attributes. Using automatically detected human attributes, they achieve excellent performance on keyword based face image retrieval and face verification. Siddiquie et al. Further extend the framework to deal with multi-attribute queries for keyword-based face image retrieval. Scheirer et al. propose a Bayesian network approach to utilize the human attributes for face identification.

## II. LITERATURE SURVEY

B.C.Chen et.al [2] developed a scalable face image retrieval system which can integrate with partial identity information to improve the retrieval result. To achieve this goal, B.C.Chen et.al first apply sparse coding on local features extracted from face images combining with inverted indexing to construct an efficient and scalable face retrieval system. Then propose a novel coding scheme that refines the representation of the original sparse coding by using identity information. Using the proposed coding scheme, face images with large intra-class variances will still be quantized into similar visual words if they share the same identity. Experimental results show that the system can achieve salient retrieval results on LFW dataset (13K faces) and outperform linear search methods using well known face recognition feature descriptors.

Attributes were recently shown to give excellent results for category recognition. A. Ramisa et al. [3] discuss their performance in the context of image retrieval; show that retrieving images of particular objects based on attribute vectors gives results comparable to the state of the art. A. Ramisa et al. demonstrate that combining attribute and Fisher vectors improves performance for retrieval of particular objects as well as categories. Implement an

efficient coding technique for compressing the combined descriptor to very small codes. Experimental results on the Holidays dataset show that our approach significantly outperforms the state of the art, even for a very compact representation of 16 bytes per image. Retrieving category images is evaluated on the “web-queries” dataset. A. Ramisa et al. show that attribute features combined with Fisher vectors improve the performance and that combined image features can supplement text features.

Huang, G. [5] discussed about the labeled faces in the wild. Face recognition has benefitted greatly from the many databases that have been produced to study it. Most of these databases have been created under controlled conditions to facilitate the study of specific parameters on the face recognition problem. These parameters include such variables as position, pose, lighting, expression, background, camera quality, occlusion, age, and gender. While there are many applications for face recognition technology in which one can control the parameters of image acquisition, there are also many applications in which the practitioner has little or no control over such parameters.

This database is provided as an aid in studying the latter, unconstrained, face recognition problem. The database represents an initial attempt to provide a set of labeled face photographs spanning the range of conditions typically encountered by people in their everyday lives. The database exhibits “natural” variability in pose, lighting, focus, resolution, facial expression, age, gender, race, accessories, make-up, occlusions, background, and photographic quality. Despite this variability, the images in the database are presented in a simple and consistent format for maximum ease of use. In addition to describing the details of the database and its acquisition, we provide specific experimental paradigms for which the database is suitable. This is done in an effort to make research performed with the database as consistent and comparable as possible.

For identity related problems, descriptive attributes can take the form of any information that helps represent an individual, including age data, describable visual attributes, and contextual data. With a rich set of descriptive attributes, it is possible to enhance the base matching accuracy of a traditional face identification system through intelligent score weighting. Factor any attribute differences between people into our match score calculation, it can deemphasize incorrect results, and ideally lift the correct matching record to a higher rank position. Naturally, the presence of all

descriptive attributes during a match instance cannot be expected, especially when considering nonbiometric context.

Retrieval-based face annotation is a promising paradigm in mining massive web facial images for automated face annotation. Such an annotation paradigm usually encounters two key challenges. The first challenge is how to efficiently retrieve a short list of most similar facial images from facial image databases, and the second challenge is how to effectively perform annotation by exploiting these similar facial images and their weak labels which are often noisy and incomplete. Wang, D. et, al. [14] mainly focus on tackling the second challenge of the retrieval-based face annotation paradigm. The propose of an effective Weak Label Regularized Local Coordinate Coding (WLRCC) technique, which exploits the local coordinate coding principle in learning sparse features, and meanwhile employs the graph-based weak label regularization principle to enhance the weak labels of the short list of similar facial images.

Recent work has shown that visual attributes are a powerful approach for applications such as recognition, image description and retrieval. However, fusing multiple attribute scores – as required during multi-attribute queries or similarity searches – presents a significant challenge. Scores from different attribute classifiers cannot be combined in a simple way; the same score for different attributes can mean different things. Scheirer, W et al, [10] show how to construct normalized “multi-attribute spaces” from raw classifier outputs, using techniques based on the statistical Extreme Value Theory. Multi attribute space method calibrates each raw score to a probability that the given attribute is present in the image. It describe how these probabilities can be fused in a simple way to perform more accurate multi attribute searches, as well as enable attribute-based similarity searches. A significant advantage of our approach is that the normalization is done after-the-fact, requiring neither modification to the attribute classification system nor ground truth attribute annotations. It demonstrate results on a large data set of nearly 2 million face images and show significant improvements over prior work.

Kumar, N. et al, [6] present two novel methods for face verification. The first method is “attribute” classifiers; it uses binary classifiers trained to recognize the presence or absence of describable aspects of visual appearance (e.g., gender, race, and

age). The second method is “simile” classifier; it removes the manual labeling required for attribute

Classification and instead learns the similarity of faces, or regions of faces, to specific reference people. Neither method requires costly, often brittle, alignment between image pairs; yet, both methods produce compact visual descriptions, and work on real-world images. Furthermore, both the attribute and simile classifiers improve on the current state-of-the-art for the LFW data set, reducing the error rates compared to the current best by 23:92% and 26:34%, respectively, and 31:68% when combined. For further testing across pose, illumination, and expression, we introduce a new data set – termed PubFig – of real-world images of public figures (celebrities and politicians) acquired from the internet. This data set is both larger (60,000 images) and deeper (300 images per individual) than existing data sets of its kind. Finally, we present an evaluation of human performance.

Victor P. et al [13] reviewed a machine learning approach for visual object detection which is capable of processing images extremely rapidly and achieving high detection rates. This work is distinguished by three key contributions. The first is the introduction of a new image representation called the “Integral Image” which allows the features used by our detector to be computed very quickly. The second is a learning algorithm, based on AdaBoost, which selects a small number of critical visual features from a larger set and yields extremely efficient classifiers. The third contribution is a method for combining increasingly more complex classifiers in a “cascade” which allows background regions of the image to be quickly discarded while spending more computation on promising object-like regions. The cascade can be viewed as an object specific focus-of-attention mechanism which unlike previous approaches provides statistical guarantees that discarded regions are unlikely to contain the object of interest. In the domain of face detection the system yields detection rates comparable to the best previous systems.

### III. PROPOSED SYSTEM

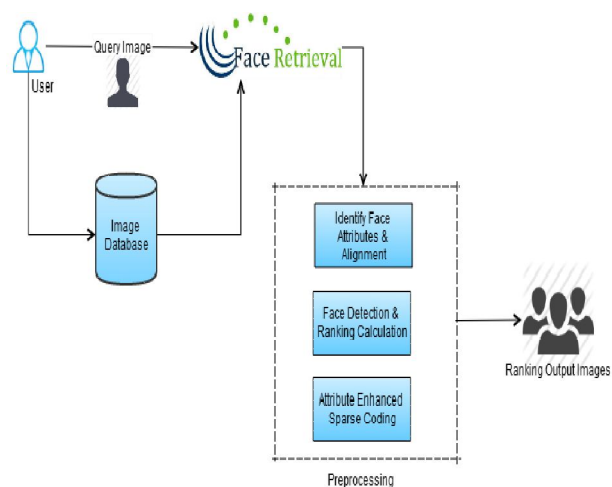
To balance global representations in image collections and locally embedded facial characteristics, we propose two orthogonal methods to utilize automatically detected human attributes to improve content-based face image retrieval under a scalable framework. Two orthogonal methods are attribute-enhanced sparse coding and attribute-embedded inverted indexing [5] which can effectively retrieve index with more than one million face photos

can be done in less than one second, reduce memory usage by many compression techniques in information retrieval, reduce the quantization error and achieve salient gains in face retrieval, improve content based face retrieval by constructing semantic code words for efficient large-scale face retrieval. It is an enabling technology for many applications including automatic face annotation, crime investigation, etc.

In proposed work, a user will upload a Query image to face retrieval module and image database will verify it to allow for preprocessing (Fig1). In preprocessing, attribute enhanced sparse coding is applied to identifies the face attributes and alignment, face detection and ranking calculation to output ranked images.

### 3.1 ATTRIBUTE-ENHANCED SPARSE CODING:

Attribute-enhanced sparse coding exploits the global



structure of feature space and uses several important human attributes combined with low-level features to construct semantic code words in the offline stage.

Fig 1: Architecture diagram

### 3.2 ATTRIBUTE-EMBEDDED INVERTED INDEXING:

Attribute-embedded inverted indexing locally considers human attributes of the designated query image in a binary signature and provides efficient retrieval in the online stage. Automatically detected high-level human attributes and low-level features are combined in the proposed work to construct semantic codewords.

To the best of our knowledge, this is the first application proposal of such combination for content-

based face image retrieval. In this application Viola-Jones face detector method is applied to find the locations of faces in the image present in the every database.

For a single query, face detection and alignment take about 0.7 seconds, computing sparse representation takes about 0.35 seconds, and retrieving index with more than one million takes about 0.02 seconds.

### **3.3 SCALABLE FACE RESTORATION:**

To complete the procedure of scalable face restoration process is enumerated below:

#### **3.3.1. DETECTING FACE ATTRIBUTES:**

The explosive growth of camera devices, people can freely take photos to capture moments of life, especially the ones accompanied with friends and family. Therefore, a better solution to organize the increasing number of personal or group photos is highly required. In this module, we propose an Attribute Enhanced Sparse Codewords to search for face images according facial attributes and face similarity of the target persons. Image retrieval systems achieve scalability by using bag-of-words representation and textual retrieval methods, but their performance degrades quickly in the face image domain, mainly because they produce visual words with low discriminative power for face images, and they ignore the special properties of the faces. The leading features for face recognition can achieve good retrieval performance, but these features are not suitable for inverted indexing as they are high-dimensional and global, thus not scalable in either computational or storage cost. In this application we build a scalable face image retrieval system. For this purpose, we develop a new scalable face representation using both local and global features.

Our goal of this application is to address one of the important and challenging problems large scale content-based face image retrieval. Given a query face image, content-based face image retrieval tries to find similar face images from a large image database. It is an enabling technology for many applications including automatic face annotation, crime investigation, etc.

#### **3.3.2. ESTIMATING FACE SIMILARITIES:**

Face image retrieval usually use low-level features to represent faces in the traditional methods, but lowlevel features are lack of semantic meanings and face images usually have high intra-class variance, so the retrieval results are unsatisfactory. To tackle this problem, we propose to use identity based quantization and also propose to use identity

constrained sparse coding, but these methods might require clean training data and massive human annotations. In this work, we provide a new perspective on contentbased face image retrieval by incorporating high-level human attributes into face image representation and index structure. Face images of different people might be very close in the low-level feature space. By combining low-level features with highlevel human attributes, we are able to find better feature representations and achieve better retrieval results.

To enable search through face appearance, we adapt the face retrieval framework. The advantage of this framework includes: Efficiency, which is achieved by using sparse representation of face image with inverted indexing, and leveraging identity information, which is done by incorporating the identity information into the optimization process for codebook construction. Both of the above two points are suitable for our system.

In details, detected faces are first aligned into canonical position, and then component based local binary patterns are extracted from the image database. Sparse representations are further computed from these feature vectors based on a learned dictionary combined with extra identity information. By incorporating such framework into our system, the user can not only specify positions and attributes of the face but also use a face image itself with position as the query. The real valued similarity scores are normalized to the interval (0, 1) before they are used.

#### **3.3.3. CONTENT BASED IMAGE SEARCH:**

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. Human attributes have been shown useful on applications related to face images; it is non-trivial to apply it in content-based face image retrieval task due to several reasons. First, human attributes only contain limited dimensions.

When there are too many people in the dataset, it loses discriminability because certain people might have similar attributes. Second, human attributes are represented as a vector of floating points. It does not work well with developing large scale indexing methods, and therefore it suffers from slow response and scalability issue when the data size is huge. To leverage promising human attributes automatically detected by attribute detectors for improving content-



based face image retrieval, we propose two orthogonal methods named attribute-enhanced sparse coding and attribute-embedded inverted indexing. In this module, we aim to utilize automatically detected human attributes that contain semantic cues of the face photos to improve content based face retrieval by constructing semantic codewords for efficient large-scale face retrieval.

### **3.3.4. ATTRIBUTE EMBEDDED INVERTED INDEXING BASED IMAGE SEARCH:**

Attribute detection has adequate quality on many different human attributes. Using these human attributes, many researchers have achieved promising results in different applications such as face verification [6], face identification [7], keyword-based face image retrieval [8], and similar attribute search [9]. Attribute-enhanced sparse coding [10] exploits the global structure of feature space and uses several important human attributes combined with low-level features to construct semantic codewords in the offline stage. On the other hand, attribute-embedded inverted indexing locally considers human attributes of the designated query image in a binary signature and provides efficient retrieval in the online stage.

By incorporating these two methods, we build a large-scale content-based face image retrieval system by taking advantages of both low level features and high-level semantics. To embed attribute information into index structure, for each image, in addition to sparse codewords computed from the facial appearance, we use a dimension binary signature to represent its human attribute. Attribute-embedded inverted index is built using the original codewords and the binary attribute signatures associated with all database images. Attribute-embedded inverted indexing further considers the local attribute signature of the query image and still ensures efficient retrieval in the online stage.

### **3.3.5. SCALABLE FACE IMAGE RETRIEVAL:**

The proposed work is a facial image retrieval model for problem of similar facial images searching and retrieval in the search space of the facial images by integrating content-based image retrieval (CBIR) techniques and face recognition techniques, with the semantic description of the facial image. The aim is to reduce the semantic gap between high level query requirement and low level facial features of the human face image such that the system can be ready to meet human nature way and needs in description and retrieval of facial image. Traditional CBIR

techniques use image content like color, texture and gradient to represent images. To deal with large scale data, mainly two kinds of indexing systems are used. Many studies have leveraged inverted indexing or hash based indexing combined with bag-of-word model (BoW) and local features like SIFT, to achieve efficient similarity search.

We can automatically detect facial attributes and measure face similarity in the offline process to provide rapid on-line photo search. Integrated with aesthetics assessment, we can further save time for browsing photos with poor quality. Using human attributes like hair colors we can gather information from not only face regions, therefore we can still achieve good performance under the occlusion. If the quality of the query image is poor, we cannot correctly predict the human attributes and sparse codewords.

### **3.3.6. RANKING IN ATTRIBUTE ENHANCED SPARSE CODING:**

In this proposed application both the query and database images will go through the some procedures including face detection, facial landmark detection, face alignment, attribute detection, and feature extraction. Attribute-Enhanced sparse coding is used to find sparse codewords of database images globally. Codewords of the query image are combined locally with binary attribute signature to traverse the attribute-embedded inverted index and derive real-time ranking results over database images. The experimental results show that using the codewords generated by the proposed coding scheme, we can reduce the quantization error and achieve salient gains in face retrieval on two public datasets; the proposed indexing scheme can be easily integrated into inverted index, thus maintaining a scalable framework.

The image ranking according to Equation can still be efficiently computed using inverted index by simply doing a XOR operation to check the hamming distance before updating the similarity scores. XOR operation is faster than updating scores, by skipping images with high hamming distance in attribute hamming space, the overall retrieval time significantly decreases. We would like to highlight what improvements we can bring in as exploiting face attributes for semantic-rich sparse codeword representations. Certain attributes (smiling, frowning, harsh lighting, etc.) will decrease the performance in both datasets. It is probably because these attributes are not correlated with the identity of the person. Informative human attributes across both datasets are also similar.

#### IV. CONCLUSION

Two orthogonal methods to utilize automatically detected human attributes to significantly improve content-based face image retrieval are used in this paper. Attribute-enhanced sparse coding exploits the global structure and uses several human attributes to construct semantic-aware code words in the offline stage. Attribute-embedded inverted indexing further considers the local attribute signature of the query image and still ensures efficient retrieval in the online stage. The experimental results show that using the code words generated by the proposed coding scheme, we can reduce the quantization error and achieve salient gains in face retrieval on two public datasets; the proposed indexing scheme can be easily integrated into inverted index, thus maintaining a scalable framework. During the experiments, we also discover certain informative attributes for face retrieval across different datasets and these attributes are also promising for other applications. Current methods treat all attributes as equal. We will investigate methods to dynamically decide the importance of the attributes and further exploit the contextual relationships between them.

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