

EAR BASED BIOMETRICS IDENTIFICATION AND DETECTION

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AbstractThe human authentication is that the important aspect for the safety requirements in very we sector. Physical characteristics of humans, like biometrics, are usually don't change over time, unique to each person and straightforward to capture. Present research in biometrics is concentrated on very us biometric iris, fingerprint, gait, signature, face, palm print, or hand-geometry. Ear is most vital part for human identification in various applications. The ear of human sort of efficient biometric candidate because it's a) Stable and appropriate structure that's preserved from birth and is exclusive in individuals b) Relatively immune from hygiene issues, privacy and biometric elements. This technique investigates single approach for the automated human identification by 2D ear imaging. Various geometrical feature extraction like identification of shape an individual using ear biometrics

IndexTerms–Ear Biometrics, Segmentation, Person Identification,

I. INTRODUCTION

Image Processing :-Image processing may be a method to convert a picture into digital form and perform some operations thereon , so as to urge an enhanced image or to extract some useful information from it. It is a kind of signal dispensation during which input is picture, like video frame or photograph and output could also be picture or characteristics related to that picture. Usually Image Processing system includes treating picture as 2D signals while applying already set signal processing methods to them. It is among rapidly increasing technologies today, with its applications in various aspects of a business. Image Processing forms core research region during the engineering and computing disciplines too.

1.1 Image processing basically includes the subsequent three steps:

Importing the image with optical scanner.

Analyzing the image which incorporates data compression and image enhancement and spotting patterns that aren't to human eyes like satellite photographs.

Output is that the final stage during which result are often altered image or report that's supported picture analysis.

1.2 Purpose of Image processing:

The purpose of image processing issplit into 5 groups. They are:

- Visualization - Observe the objects that aren't visible.
- Image sharpening and restoration - to make a far better image
- Image retrieval - search for the image of interest.
- Measurement of pattern – Measures various objects in a picture.
- Image Recognition – Distinguish the objects in a picture.

2. Literature Survey

2.1 On the utilization of external Ear Images for private Identification in Security Applications

Moreno, B., Sánchez, Á. Vélez. J.F. In this paper we investigate the utilization of external ear images for human identification. From the purpose of view of image processing, ears offer several advantages over complete faces: reduced spatial resolution, a more uniform distribution of color, and fewer variability with expressions and orientation of the face. These advantages along side its identification richness, make ear images appropriateto be used as input file for a connectionist system. a replacement multiple identification method, which mixes the results obtained by several neural classifiers using, respectively, features external ear points, information obtained from ear shape and wrinkles, and macrofeatures extracted by a compression network, is presented. Experimental results yields higher identification rates also as a more robust framework using this approach as a component of a more general face identification system especially in security applications.

2.2 Automated human identification using ear imaging

A. Kumar, C. Wuthey investigates a replacement approach for the automated human identification using 2D ear imaging. We present a totally automated approach for the robust segmentation of curved region of interest using morphological operators and Fourier descriptors. We also investigate new feature extraction approach for ear identification using localized orientation information and also examine local gray-level phase information using complex Gabor filters. Our investigation develops a computationally attractive and effective alternative to characterize the automatically segmented ear images employing a pair of log-Gabor filters. The experimental results achieve average rank-one recognition accuracy of 96.27% and 95.93%, respectively, on the publicly available database of 125 and 221 subjects. Our experimental results from the authentication experiments and wrong identification verses wrong identification also suggest prevalence superiority of the proposed approach over the opposite popular feature extraction approach considered during this work.

2.3 Ear Biometrics, "Biometrics: personal identification in networked society"

M. Burge, W. Burger Personal Identification in Networked Society may be a comprehensive and accessible source of state-of-the-art information on all existing and emerging biometrics: the science of automatically identifying individuals supported their physiological or behavioral characteristics. "Biometrics: Personal Identification in Networked Society is a useful work for scientists, engineers, application developers, systems integrators, et al. working in biometric.

3. RESEARCH METHODOLOGY

Proposed method follow process related with the ear detection techniques is basic methodology for the identification of ear images. Input ear image taken by the AMI ear database. Preprocessing which is employed to seek out out the sting detection using various techniques like canny, these are the some samples of the sting detection. Following are the steps of proposed system:

- Image Pre-processing
- Region of Interest Detection
- Feature Extraction
- Identification

The detailed information about these steps is given below.

3.1 Image Acquisition:

Image acquisition is first stage of image processing applications. So as to check and compare the detection or recognition performance of a computer vision system. generally and a biometric system especially, Image database of sufficient size must be publicly available, therein we used the datasets. The dataset contains images from the left and therefore the right ear subjects. the pictures were taken under varying lighting conditions and therefore the subjects weren't asked to get read of hair, jewelry or head dress before taking the photographs. the picture are cropped from video stream, which shows the topic in several poses, like looking towards the camera, upwards or downwards. Additionally, the bottom truth for the ear's position is provided along side the database, which makes it particularly convenient to get the accuracy of ear detection and to review the ear recognition performance independently from each ear detection.

Preprocessing involves converting the image to gray scale, performing histogram equalization, and Gaussian filtering. Pre-processing is important so as to get rid of remove noise and smooth the image.

• **Converting to Grey-scale** involves mapping color RGB triplets to a one value representing the grey scale intensity. every color pixel is described by a triple (R, G, B) of intensities for red, green, and blue. A weighted average of these values is calculated because the grey scale intensity as follows:

$$I = 0.21 R + 0.72 G + 0.07 B$$

A weighted average is employed for human perception, because humans are more sensitive to green than other colors, therefore green carries the most important weight.

3.2 Preprocessing - Cropping of ear and Normalization

Before using the raw depth scans of the profile picture, some preprocessing steps are to be performed. This preprocessing wont to improve the visual appearance of a picture. If any distortions are present remove this stage. And if any noise scans we apply Imfilter Gaussian filter and to get rid the noise. during this preprocessing are convert the colour image into gray scale image.

3.2.1 Region of interest detection:

Region of interest detection involves identifying the boundary of the ear within the image and extracting it. To implement the region of interest we will use the Hair Feature-based Cascade Classifier.

• **Hear Feature-based Cascade Classifiers** is trained with a couple of hundred samples of object. In this case ear images are called positive images and arbitrary images of an equivalent size, called negative images. Then the classifier is applied to a picture and outputs as 1 or 0 are obtained counting on whether the image contains the things of interest (in this case, the ear) par not. This approach isn't 100% effective, and in many cases, the first image has got to be used.

3.3 Segmentation -Ear detection using Edge detection method and Morphological operation.

Binary images may contain countless defects. In some circumstances binary regions constructed by noise and textures. Morphology may be a vast extent of image processing that modifies the pictures supported operations shapes. it's considered to be one amount the info processing methods useful in image processing. it's has more applications such as texture analysis, noise elimination, boundary extraction etc. Morphological image processing shows the goal of eliminating of these defects and maintaining structure of picture. Morphological operations are confident only on the associated ordering of pixel values, instead of their numerical values, in order that they are focused more on binary images, but it also can be applied to gray-scale images such their light transfer functions are unknown and thus their absolute pixel values aren't taken into consideration. Morphological techniques verify the image with a little template called structuring element. This structuring element is applied to all or any possible locations of the input image and generates an equivalent size output. during this technique the output image pixel values are supported same pixels of input image with is neighbors. This operation produces a replacement binary image during which if test is successful it will have non-zero pixel value at that location within the input image. There are many structuring element such as diamond shaped, square shaped, cross shaped etc. The bottom of the morphological operation is dilation, erosion, opening, closing expressed in logical AND, OR notation and described by group analysis. Among them during this paper only two operations are used dilation and erosion. Dilation adds pixels while erosion removes the pixels at border of the objects. This removal of pixels depends on the structuring element used for processing the image.

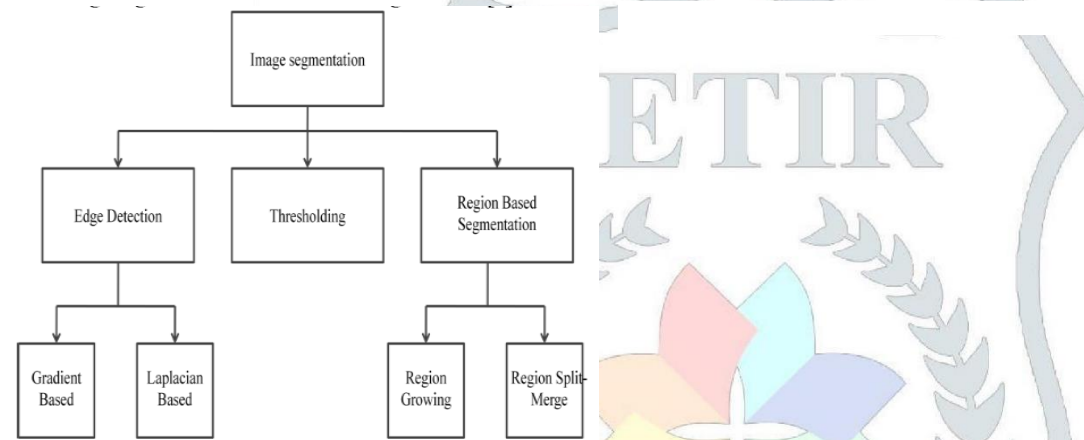


Fig 1: Overview of image processing

3.4 Feature Extraction - From edge detected ear.

The process next to tracking is feature extraction. the most purpose of feature extraction is data reducing by measuring certain features that distinguish objects or their parts. during this feature extraction we use the feature score supported selection of contour of an ear. during this stage longest distance points decided. For this distance matrix for every contour point was computed. the 2 points on the contour which are farthest apart are selected. These two points span the most axis of the ear feature vector following the middle point on the road is calculated relative two that line, 3 points are calculated. Supported these points the geometric feature value is calculated. Concluded that geometrical features representing shapes of ear are large suitable for ear picture than texture, color or global features. Our method is split into various image normalization, extraction, calculation of the centroid, coordinates normalization and a couple of steps of geometrical feature extraction, as described within the next section. We treat the centroid because the specific point in our method, albeit it's not a selected point within the ear topology.

There are various possible geometrical technique of feature extraction and shape description such as Fourier Descriptors, methods supported combination of angles and distances as parameters. We propose a 2 step-method that's supported number of pixels that have an equivalent radius during a circle with the

Center in the centroid and on the contour topology. The algorithm for the primary step of feature extraction is Presented below:

1. We create a group of circles with the middle within the centroid
2. Number of circles N is fixed
3. We create circles in such a fashion that the corresponding radiuses are α pixels longer from the previous radius
4. Since every circle is crossed by the contour image pixels we count the amount of intersection pixels r
5. Then we calculate every distances d between neighboring pixels, we worked after the counterclockwise direction
6. We build the feature vector that consists of all the radius with the corresponding number of pixels belonging to every radius and with the sum of all the distances between those pixels $\sum d$

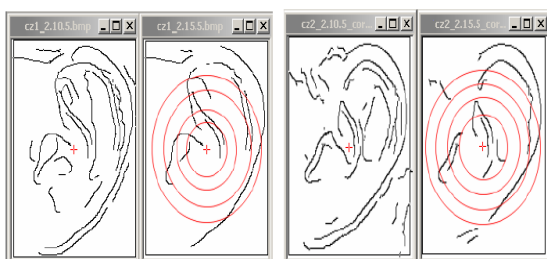


Fig 2: Binary ear images with the extracted edges (2 values of k) and with the centroid denoted with a cross. Circles denote the radius values for calculation of the number of pixels intersecting each of circle. The table below shows the centroid values for every binary image.

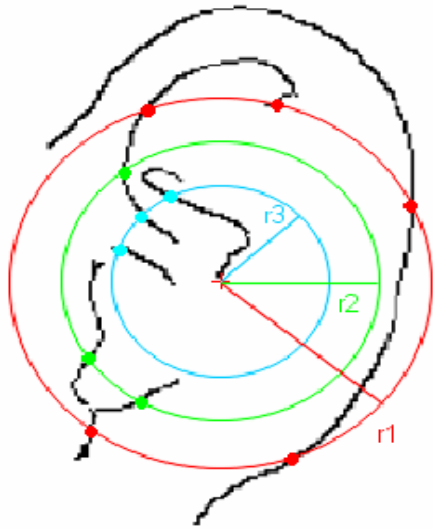


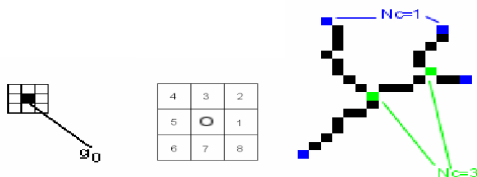
Fig 3: Matching points using surface
 The algorithm for $r = 3$ is symbolically presented in the Figure
 Fig. The symbol of our algorithm for $r = 3$ is N
 The general rule for forming the primary vector is presented below:

$$V = \left\{ \left[r_{\min}, l_{r_{\min}}, \sum d_{r_{\min}} \right] \cdots \left[r_{\max}, l_{r_{\max}}, \sum d_{r_{\max}} \right] \right\}$$

where:
 r – radius length,
 l – number of intersection points for every radius,
 $\sum d$ – sum of all the distances between the intersection points for the considered radius.
 Then so as to reinforce the distinctiveness of the extracted features, we build the second vector within the second step of feature extraction. Once again we base upon the created circles with the middle within the centroid.
 For each contour line the characteristic points are:
 - contour endings,
 - contour bifurcations,
 - every points that cross that created circles (those points are already extracted by the previous algorithm).
 In each contour we check the topological properties of each pixel. For every contour pixel g we use 3×3 window as in Fig (left).
 When $g = 8$ is the connected number
 N of g is defined as:

$$N_c^s(g_o) = \sum_{k \in S} \left(\bar{g}_k - \bar{g}_{k-1} - \bar{g}_{k+1} - \bar{g}_{k+2} \right),$$

where $S = (1,3,5,7)$ and $\bar{g}_k = g_k - 1$



Feature extraction deals with isolating distinct features of the ear within the image. we will use various techniques like PCA, LBP and spatial histograms to accomplish this process.

3.4.1 Principal Component Analysis (PCA) may be a method of dimensionality reduction, which deals with the reduction of the amount of features to only those with an outsized variation between them. Firstly, each pixel in a picture is taken row by row and converted to a row vector of the intensity values. All the row vectors of the training set (or testing set) are combined to make a matrix. Principal components analysis (PCA) is an orthogonal linear transformation that transforms the info to a replacement frame

of reference such the best variance by any projection of the info involves lie on the primary coordinate (called the primary principal component), the second greatest variance on the second coordinate, and so on.

3.4.2 Local Binary Patterns (LBP) deals with comparing the pixel value with its neighboring pixels values. It doesn't check out the image as an entire, but instead isolates local features of an object. Each pixel is compared with only its neighborhood. A pixel is taken because the center and used as a threshold value. If the intensity of the neighbor pixel is bigger than or equal the middle pixel, then denote it with 1. Otherwise it's denoted with a 0. The result's a binary number for every pixel, eg.11001111, called an LBP code. Local Binary Patterns reveal the feel of a picture, and therefore the features extracted have a coffee dimensionality.

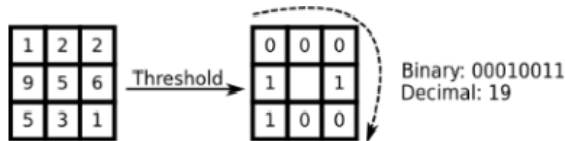


Fig 4: Example calculation of LBP

Spatial Histograms are wont to preserve local information in a picture. rather than computing one histogram for the entire image, the image is split into $n \times n$ smaller regions, and a histogram is calculated for that every region separately. The region size used was 8×8 .

4. Conclusion

We have developed a totally automated approach for identification of an individual. We've taken the ear as a biometric candidate because ear possesses properties which make it unique and therefore the best candidate. We use geometric approach and use feature extraction to extract a neighborhood of the ear. Feature extraction is performed by using cascade classifiers. once the ear is detected the picture of the ear is stored in database in blob format. For authenticating an individual image comparison is completed. We make use of histogram generated of the pictures for comparison purposes. The results obtained are reliable and accurate. Hence a strong and reliable system is developed which may be utilized in various countries for human identification and authentication. The system developed is platform independent and hence it are often used on multiple platforms.

3. References

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