IRIS Recognition Segmentation Signal Processing System

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ABSTRACT

Over the past ten years, the growing emphasis on security has drawn a lot of interest in iris recognition in both academic and practical settings. Due to its high accuracy, reliability, and uniqueness, iris recognition is becoming increasingly popular in the fields of access control, electronic commerce, and border security. The iris is the best potent identification and verification feature when compared to extra biometric modalities because the human iris does not change throughout the course of a person's lifetime. Feature extraction, segmentation, template matching and normalization are the first four steps of iris recognition. These preprocessing techniques used by different researchers for iris segmentation are the focus of this review paper.

Keywords: active contour technique, iris recognition, segmentation, canny edge detection

I. INTRODUCTION

For example, biometric identification method uses some kind of unique feature or characteristic to identify an individual. Biometric identification based on iris patterns is called "iris recognition" and is a form of automatic biometric identification. Given that it possesses distinguishing qualities such as furrows, rings, ridges, freckles, and a high degree of randomness, detailed patterns, the need for Iris recognition is increasing daily in several access control industries, security at border areas, and so on. Iris patterns differ from person to person. Both eyes of a person are unique, even if they are identical twins or identical siblings. Human identification can be highly reliable and accurate due to the uniqueness, stability an individual's lifetime entire life, inability to forge, and the protection provided by iris recognition.

In 1987, scientists Flom and Safir proposed using iris recognition as a biometric. The iris recognition field was flooded with new algorithms in the following years, including those developed by researchers a like K.W. Bowyer, John Daugman, A.K. and Jain Wildes,. The iris is the part of the eye that is visible to the outside world. Image segmentation, normalisation, feature extraction, and template matching are the steps in iris recognition as depicted in Fig. 1.

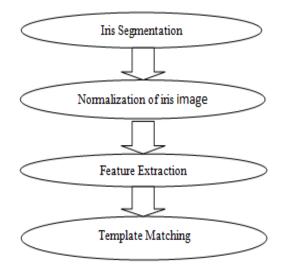


Figure 1: The steps of iris recognition

II. ASSESSMENT OF IRIS

After acquiring an iris image, the next step in iris recognition is image segmentation. The iris is removed from the digital image of the eye. If you want to generate templates, you need to know where the iris's outer and inner boundaries are located. Researchers have used a variety of methods to segment the iris. Here are a few examples:

A. Conventional Methods

There are many edge-detecting software programmes, such as Canny Edge Detector and Hough Transform used in these methods of iris segmentation to identify Iris borders. Although the outer and inner boundaries of an iris appear to be circular, this is not the case, and they do not segment the region accordingly. Following are a few examples of traditional iris localization methods:-

2.1 Integro Differential Operator

Iris recognition systems frequently employ this Daugman-developed segmentation technique. Iris boundaries and upper and lower eyelids can be identified using this technique, which uses the maximum value of the contour integral along concentric circles of smoothed radial image derivative to compute the contour integral. Faster computations are possible because it makes use of the first derivative of the input data. The Integro differential operator is given by Equation(1).

$$max_{(r,x_0,y_0)}|G_{\sigma}(r)*\frac{\partial}{\partial r}\oint r, x_0, y_0\frac{l(x,y)}{2\pi r}ds|$$
(1)

I(x, y) is an image, ds is a circular arc with radius r, (x0, y0) is the centre coordinate, * stands for G (r) is a function for smoothing and convolution, respectively.

2.2 Canny's Edge Detection Tool

It's a tool for spotting iris borders. It reveals the eye's major edges, making it simple to pinpoint the iris boundary. It is used to clean up the image and get rid of any fuzziness. Errors can be found using this method. In spite of the fact that it performs complex calculations, it takes less time and less memory space to run. It's a very straightforward approach.

2.3 Circular Hough Transform

A strong algorithm and common computer vision can be used to infer the follower and iris areas. The control chart is the desired result; therefore, thresholding is used after computing the initial byproducts of the strength values in an eye picture. An equation may define any circle by taking into account factors like the Xc and Yc coordinates for its centers and the radius (r) that runs across each of its edge points.

$$X_c + Y_c - r^2 = 0 (2)$$

If there are many calculations, the processing pace is slow [10].

2.4 K-means Clustering

Using this technique, the eye image is divided into three distinct areas based on their intensity. The iris, which includes the pupil's eyelashes, is represented in the first region. The sclera and luminance reflections in the second region have high intensity values. In between these two regions and on top of the skin is the third region. Because the upper eyelid is occluded by an arc, the iris does not lose any of its useful areas.

2.5 Active Contour Models

These contours are treated as deformable borders in this algorithm (not circular shape). Iris localization from the sclera and the pupil is improved. Listed below are the active contouring methods. The foundation of traditional edge detection techniques is the evolution of a curve around an object after an initial shape has been bent toward the object's border. The energy function for minimizing an active contour model is made by combining, in a weighted way, the forces that are created by the snake shape and the forces that are created by the image.

The equation provides the definition of the energy function (3):

$$Esnake = \int_0^1 E_{snake} (V(s)) ds \qquad (3)$$

$$Esnake = \int_0^1 E_{int} (V(s)) + E_{image} (V(s)) + (V(s)) ds$$

Where

 $E_{int}(V(s))$ indicates the spline's internal energy as a result of bending. $E_{imag}(V(s))$ It depicts the forces of the image. $E_{con}(V(s))$ It displays the external constraining force.

2.6 Edgeless Active Contour

On the boundaries of discrete gradients, the halting function never equals zero, so the curve may pass through the boundary. To address this issue, a new active contour model has been developed: the Active contour without edges model. Eyelashes and corneal reflections have no effect on this model's ability to accurately locate the pupil.

2.7 Gradient Vector Flow Snake

It is hardest to pinpoint the iris' outer margin, compared to the inner iris boundary. Because of the lack of intensity difference between the iris and sclera boundary, Without an accurate edge model, it cannot be divided into segments by active contour. The Gradient Vector Flow Field is defined by GVF Snake as a new external force field to address this issue.

2.8 Statistical Learning Methods

The edges, the coarse iris's centre, and its radius, and the delicate iris borders and centres are planned with the square with the lowest median and three-step iris segmentation using a linear basis function. In order to reject low-quality images, such as those that are out of focus or fail to segment, it focuses on image quality evaluation.

2.9 Least Median of Square Differential Operator (LMSDO)

We use this procedure to determine the inner and outer borders of the iris's coarse centres and radiuses. The cost of this increased calculation is that it is less reliable than the IDO (Integra Differential Operator). An equation provides the answer (4)

$$\max_{(r,x,y)} |G_{\sigma}(r) * \frac{\partial}{\partial r} med\left(I_{r,x,y}(x,y)\right) ds |$$
(4)

Where med stands for median **2. Linear Basis Function (LBF)**

It is employed to determine the iris's rough inner and outer edges. It is favoured because it offers quick and flexible computation by using the trigonometric works ($\cos(2\pi K / N)$, $\sin(2\pi K / N)$), with (K = 1, 2) as the foundation works. The equation provides the information. (5)

$$y_i = W^T \varphi(n_i)$$
 (5)
Where $i=1,2,3...,N,W=(W_0,W_1,...,W_4)^T, \phi(n_i)=(\phi 0...,\phi 4)^T$

III. RANSAC (RANDOM SAMPLE CONSENSUS)

Occlusion by eyelids, eyelashes, and speculations is a common cause of out-of-focus points in an image. The following are the components of this procedure:

- All of the coarse boundary points are subject to the Linear Basis Function perfect.
- We reject the limit facts that are too far from our coarse boundary points because we consider them to be outliers.
- All the remaining coarse boundary points are modeled using a linear basis function.
- Go back to step 1 and notice that a linear basis function is used to model each coarse boundary point.

As a loop, this procedure can be repeated as many times as necessary. These inner and outer curves are used to calculate the fine iris' centers and are denoted as (xi,yi). The fine centers are the weight centers of the inside and outer limitations. It can be found in the equation (6) (7):

$$x_c = \frac{\sum_{i=1}^N x_i}{N}$$
(6)
$$y_c = \frac{\sum_{i=1}^N y_i}{N}$$
(7)

Where (x_c, y_c) are iris centres, (x_i, y_i) are the original iris image's tiny border curves used to determine the coordinates of final boundary points.

IV. EDGE DETECTION WITH RADIAL SUPPRESSION

The inner and exterior limits of the distorted iris are located using a non-separable wavelet transform. With 99.75 percent accuracy, , which is better than previous methods, the circular overpowering control detection control maps eliminate Eyelash's iris texture and edges and the boundaries of the iris can now be accurately located regardless of the pupil's shape, be it noncircular or elliptical. The edges that aren't connected to one another are then removed using the thresholding procedure.

V. CONCLUSION

Various techniques for iris segmentation have been proposed by various researchers over time, with varying levels of segmentation accuracy. The long-term stability of iris makes it an ideal material for use in places like airports, harbors, and research labs. Iris recognition has gained a lot of popularity recently due to its efficacy and accuracy.

REFERENCES

- 1. S.Prabhakar, S.Pankanti, & A.K Jain. (2003). Biometric recognition: Security and privacy concerns. *IEEE Security and Privacy*, pp. 33-42.
- 2. Khalid A. Darabkh, Raed T. AI-Zubi, Mariam T. Jaludi, & Hind AI-Kurdi. (2014). An efficient method for feature extraction of human iris patterns. *IEEE*.
- 3. Shaabad A.Sahmoud, & Ibrahim S. Abuhaiba. 92013). An efficient iris segmentation method in unconstrained environments. *Pattern Recognition*, 46(12), 3174-3185.
- 4. Richard P. Wildes. (1997). Iris recognition: An emerging biometric technology. IEEE, 85(9).
- 5. K.W. Bowyer, K. Hollingsworth, & P.J. Flynn. (2008). Image understanding for iris biometrics: A survey. *Computer Vision Image Understanding*, *110*(2), 281-307.