

# An Unsupervised Iris Based Biometric System with Inherent Feature Thresholding

Venkata Rathnam Korukonda, E. Sreenivasa Reddy



**Abstract:** This paper provides a general indication of the existing approaches rely on basic factor (i.e. extraction of iris information, affine transform, and distance matrix) as input. An essential factor in effective design of IRIS based biometric approach is the accuracy with which the model can estimate a region of interest (i.e. IRIS) within constraints and unforeseen issues, which can be very problematic but need of the hour. We introduced a new IRIS based biometric system that incorporates various factors that takes complete information of eye for developing the feature set (digest) while affine transforms are not incorporated while the three sets distance measure is incepted to enhance accuracy. The algorithm-based size of template, functionality of distance measure and/or scope, methods and/or function of application through well-defined scientific and statistical principles. Unfortunately, the accuracy of the existing approaches is limited despite the large scale of experience with several improvements based on digital image processing and statistical models. Henceforth, we incorporated several texture analysis algorithms with computing techniques along with several parametric enhancement constraints to ensure the feasibility, effectiveness and efficiency of the proposed framework in comparison with the existing methods.

**Keywords :** Iris Biometric, Texture Analysis, Applied Statistics, Wavelet Transforms, Image Enhancement.

## I. INTRODUCTION

Biometric based engineering has become the most popular engineering discipline within a short panel of the time that could address several authentication-based systems. Most of the world is dependent upon authentication-based systems for approving the integrity of the individual either for access control, ownership authentication, etc. The set of biometric system frame work activities consists of various tasks like requirement elicitation, feasibility exercise, estimation, modeling, designing, development, quality checking with different measures, and debug plans [1-5]. There is a serious need to think much about the situation in order to overcome the challenges arise due to conflicting demands within the considered parameters and limitations. To deal with the above challenges proper care should be taken while preparing the identification, processing and verification/authentication process.

Revised Manuscript Received on January 30, 2020.

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In this paper, we propose an innovative IRIS based biometric model which should aid the individual to develop authenticate systems which reduced digest size which ultimately minimizes the false error rate. The inexactness of the development of biometric system and overall verifying a wrong person or not verifying the right person were the main causes for the algorithms proposed has long been a source of frustration for many organizations [6]. In recent years, various frameworks were introduced for biometric authentication; unfortunately, their estimation accuracy is very low which includes algorithmic models. The excellence and scope of any biometric based approaches doesn't limit to the identification, processing and verification/authentication process which is considered, but based on the possible estimates of biometric image that needs to converge to the actual one. Texture analysis is one of the vital process for effective identification whose estimate parameters can be designed that are adaptive to the image, for enhancement operation and digest transformation process. It is evident that more than one approach of image process operations is essential so that there is a meaningful resource available for the estimation of the accurate biometric digest that is critical for unique application. It is evident fact that there is no clear definition of the process which could calculate and perform biometric cost benefit analysis without some logically accurate approach of identification, processing and verification/authentication process for biometrics system [7-8].

In addition, the parameters employed in calculation are sensitivity to different biometrics, image processing operations, wavelets and distance measure factors. Biometrics-based engineering illustrates the basic need of good engineering techniques such digital image processing operations (from edge detection to enhancement operations) which also acts as an important part of the base for biometric methodologies [9]. For given application, the biometric systems can be based on the parameters such as the hardware, environment factors, human resources and essential executable codes. The major portion of the biometrics development effectiveness is associated with the biometric resources (i.e. fingerprints, face and/or iris) needed which is focused by the existing biometrics systems. Further, the development are proportional based parameters of the biometric features and the nature of the application it is being employed, the precise feature extraction of the biometric largely depends on the quality of the information that is available about the biometric and after enhancement outputs.

Users must specify the process of system in a well-defined manner and correctly and uncertain parameters are condensed and precisely enhanced for the generation of biometric assessments and verification process [10].

# An Unsupervised Iris Based Biometric System with Inherent Feature Thresholding

Despite several short comings, the biometric based approaches have evolved drastically and gave precise authentication results. The primary steps in any biometric process were to comprehend, evaluate, analyze and characterize the biometric system based on the requirement and application in which it would be incorporated. In case of biometric, it is subtle, imperceptible, and inflexible making it essentially a difficult task to recognize and approximate a biometric information or proposed method that can't be shown wrong and touched the initial aspects but analyzed based on the outcome as it changes with reference to code is drafted [26, 37]. These methods available are classified into non-algorithm estimation, expert team experience-based estimation, algorithmic methods such as fingerprint method, face method, iris method and others [11]. Based on the simulation results, we observe that each of existing approach has advantages as well as drawback in comparison with other approaches, as their advantages and limitations are often complimentary to each other. The major issues of not estimating a precise biometric process can be credited to a several reasons of which the prime focus of this exposition is followed by below things

- Establish a methodology and approach for precise IRIS biometric based on algorithmic and non-algorithmic factor
- Analysing existing IRIS based biometric approaches and establishing the diversity between the existing method with respect to image processing operations
- Initiate accurate IRIS focussed based techniques for IRIS focussed analysis.
- Identify with the basics of IRIS biometrics and reason IRIS may not be directly related to its development.

Recent developments in IRIS and image processing-based technologies marked a new chapter in the design and development of IRIS based on texture analysis wherein the major problem lies in the mode of IRIS analysis (sub-blocks overlapping, image enhancement, and wavelet transform) as we employ frequency and time domain technology. Due to the IRIS based texture analysis are always displacing the existing techniques from their positions which makes it even more difficult to analyze while enhancing possible accuracy with respect to IRIS images. IRIS based biometric analysis model has two open issues that need to be understand and expressed in simple expression that could be addressed in engineering domain. It is essential to comprehend the holistic capability of the organization factors to offer the expected solution within specified schedule and environment. Furthermore, it should be able to precisely forecast the biometric analysis while the need to present the biometric application based on IRIS else the entire assessment will be hauled. The major issue that defines the schedule of image quality depends largely on luminance and the precise extraction of IRIS region (ROI) through image processing process and the corresponding Enhancement and wavelet transforms formed based on the biometric system in consideration. It is well documented based on the experience researchers that initial assessment is crude form wherein the estimates are proposed based on risk factors and prior assessments experience which cannot be assumed as the essential information for estimates. Therefore, there must be a meticulous procedure that permits a detail analysis and considers other parameters within the problem domain. An effective estimating model factors together three primary components for a precise accurate cost assessment of

the project they are region of interest (IRIS region), complexity metrics, and possible/experienced risk factors.

## II. LITERATURE SURVEY

In past few decades, relatively very little head way was made within the field of IRIS based biometric system with prime focus on texture analysis techniques, while the time and frequency domain analysis associated with were becoming vital and increasing exponentially to numerous large employing factors. Now-a-days, IRIS based biometric system and analysis are the most significant issues that influence biometric based assessment process. Most of existing approaches relates these metrics that basis of the texture analysis, but the precise figure of biometric digest can only be determined after the completion of the entire image processing operations. Henceforth, the precise assessment of the biometric digest in consideration becomes significantly more difficult task. Most of the existing approaches are broadly divided into two major classes based on the mathematical expression employed i.e.

1. non-algorithmic and
2. algorithmic.

The algorithmic class-based approaches differ mostly based on the mathematical superiority that are simple arithmetic notations which formulae using summary of statistics [16]. While some of the models employ a complex expression based of differential equations [17-19] and mathematical regression-based approaches are also formulated in last few years [23-28]. Biometric verification and detection include detection of biometric, tracking of active and inactive templates and assessment through data comparison and processing. To enhance accuracy of the biometrics, the dynamic multi-modal biometrics-based applications systems framework dynamically couples image processing with statistical models to enhance system performance. Due to heterogeneity nature of biometrics from person to person, it is very challenging to build an efficient biometric authentication system, which could either be a performance bottleneck or the single point of failure. Henceforth due to complexity of IRIS biometric and image processing technology, this chapter introduces a theoretical overview of different techniques that incorporate a decentralized area selection within the IRIS for template formulation and authentication to enable effective protection. To achieve secure identity authentication, the image processing leverages the texture detection to create virtual trust blocks, in which distributed components could identified and update the template.

To reduce error rate and other security risks such as false positive detection or true negative detection, and it requires that only authenticated features and tabulated entities of the IRIS can form the template information and use these metrics for identification system. The conventional IRIS texture access control approaches have been widely used in the IRIS biometric systems use a combination of segmentation and enhancement to identify texture region within the eye captured [16]. However, the existing solutions are not fully adapted to image processing ecosystem due to the constrained resources of space and frequency transformation of the image processing.

The combination of multiple approaches and technologies can lead to a solution of improved accuracy, smaller template and speedy detection IRIS biometric system. Furthermore, today's access control solutions often rely on centralized architecture, which not only demonstrates enormous scalability issues in a distributed environment composed of large number of templates, but also can be a performance bottleneck or the single point of failure. These approaches estimate the biometric digest based on the analogy with one or more similar concepts based on the affine transform along with supportive reasoning. This analogy is generally employed at modular level of the process (i.e. at each stage) or it can also be used to assess the entire biometric system each having its own advantages and corresponding limitations. The entire biometric system would consider the individual valuation of all functions of the system along with their linkage process. The modular based models offer a detailed analysis based on the similar functionalities and biometric digests between proposed and the existing systems [22]. There are several others method that estimates the iris region based on supervising extraction wherein the existing individual/expert offer an optimized estimate based on corresponding approaches, analysis, constraints and experience.

The main objective of IRIS based biometric development is to compute the logical and reliable area for IRIS for analysis, but so many complications are to happen when we are computing consistent measures for different approaches. To reduce this type of problems, the we employ localization and globalization concepts that defines a menu which tells about defining the inherent measure by using logical feature set while global defines the outlining parameters that could be discarded as noise. Several researchers have designed descriptive analysis, prescribed equations, additional measures to assist these computing definitions. Before starts of the IRIS based biometric system design it is vital to estimate the iris size of the eye in terms of sub-blocks exactly as possible, it can be feasible by using texture analysis. But unfortunately, how many sub-blocks required for the current proposed approach is completely based on biometric digest and image enhancement operation. It is evident that most existing estimation strategies are focused primarily of the defining the overlapping sub-block size, but it highly impossible to predict the size of the biometric digest at initial development stages with accuracy. Thus, the assessing approaches were designed which could exploit the metrics associated with size of biometric digest.

**Research Analysis & Gaps:** These traditional approaches cannot address the existing need of real-time monitoring and verification between the biometric of person and personnel therefore deployment of IRIS based biometric system is gaining the momentum even though it requires high computational complexity units. With the raid developments within the field of image processing technology the biometric authentication systems based on IRIS is still in budding phase due to the unique characteristics such as capturing capacity and variable delays such exposure of eye etc. To achieve this objective, sensors and devices needs to have a self-organize capability to make the system autonomous that can be adaptive in nature and then we can address wide range of applications.

**Localization of IRIS information (Detection):** This initial stage of the recognition system i.e. segmentation of

IRIS region which is considered as a vital aspect of the system as the accuracy highly depends on how well the IRIS could be segmented. There are several kinds of external factors that affect this process (namely eyelids, eyelashes, lamination-reflections and pupil). Henceforth most of the work focuses on the identification of circular boundaries of both quadrants to limit system to focus on IRIS ROI only through Hough transform and edge detection. We have found in most cases the prediction of the boundaries has very often several errors and valuable information needed for recognition is lost which addresses which is still an open area [43-44]. We found that information of IRIS needs to maximum available it is structure and pixel distribution from other areas is different enabling us to incorporate overlapping 3x3 block-level technique to cover image as select the all possible pixels pertaining to IRIS ROI. This process would ensure we have complete information of the IRIS

**Extraction of features from IRIS information:** This is the second stage of the recognition system i.e. analysis of the IRIS region which also considered as a vital stage of the system as the template formulation is carried out in this stage and the accuracy highly depends on how well the IRIS could be analyzed. There are several kinds of process available that could be incorporated to attain the desired results (namely enhancement, filtering, time and frequency analysis via transforms and etc.). Henceforth most of the work focuses on the conversion of the circular IRIS information into rectangular template with needed buffering to ensure the size if the fixed. We have found in most cases that the information is normalized, filtered and enhanced to ensure all the data is available in the needed format for extraction of changes in the formulation of template. The changes are analyzed based on the transforms (Gabor and others), wavelets, and image processing operations which have necessary information needed for template formulation [45-47].

### III. IMAGE PROCESSING ANALYSIS: IRIS

The rapid evolving technologies fueled with image processing has gained momentum within the field is a rapidly growing area in biometric based applications. The art of measuring the physical and physiological behavior traits that are often manually verified traditionally are now switched to digital platforms making their applications more viable and affordable.

In layman terminology, digital image processing in perspective with biometrics is concerned primarily with analyzing the images to extract needful information to create a digest (template of the biometric) which is currently processed by the computers with minimal to nil intervention from humans.

The accuracy, the variability, the quality of features and the effect intervals of the estimate biometric measures are analyzed. Furthermore, there fusion of the measures unit could offer higher results that are brought out by reducing variety of parameters in the constant models. The fundamental challenges in incorporating the digital image processing operations over IRIS images for analysis IRIS ROI and time needed for processed as an outcome [3].

## A. Biometric: IRIS

IRIS based recognition system like most of the biometric systems/approaches comprises of three major steps i.e. data acquisition, data normalization, and data matching wherein each step has its own internal phases that are defined based on the application and IRIS image quality. It also removes the danger within the IRIS images such as eye lashes overlap, lighting conditions, natural noise induced by image capturing and iris coverage. A reasonably precise assessment on the processing under various heads is vital to ensure the feasibility of a given enhancement for an application to both the verification and validation. This process limits most of the wrong estimates and preserves that biometric information on deliberations between image processing to matching operations. It is evident that more than one algorithmic approach of biometric feature estimation is essential so that there is a meaningful resource available for the estimation of the accurate actual verification and validation statement that is critical for unique application and matching operation. Considering the above limitations, the important question arises when the assessment high depends on the iris segmented that could not be precisely estimated by any method then how to optimize the difference between actual and segmented assessment of the entire iris. This motivated us in extensive research to the address the above problem by deriving an expression iris estimate for a given application based on the image processing experience and segmentation assessment via texture detection. In this paper, we introduce an innovative framework to develop a texture-based assessment technique that could effectively and efficiently address the issues while enhancing the existing processing with consideration of attributes from IRIS based approaches.

## B. Texture Framework

This process involves the different level of framework input on information network of pixels from non-overlapping blocks where user uses to ask so many questions regarding the perfect region of interest. It assimilates these inputs and constraints with reference to ROI eye ball, IRIS and white area it derives meaningful information by means texture analysis combined with entropy. The basic of the texture framework with estimation profile is discussed along with the assures of the approach in terms of the following aspects: 1.) It would efficiently assess the eye ball of each iris to pixel level of the IRIS independently based on 9-local neighborhood, 2.) It would ensure quality assessment with IRIS incorporates all the sub-blocks with overlapping/non-overlapping deliberations from entropy to mean to variance, 3.) It would incorporate a dynamic factor to estimate the unforeseen and foreseen outliers during execution of the segmentation process and 4.) It would present a detail scheduling and corresponding biometric-signature assessment from the IRIS image through a sequence of image processing operations in consideration. The texture framework developed could produce a dynamic IRIS assessment model that could predict the iris in precise and consistent manner but also provides a means for better estimation of the eye information. The statistical class-based approaches are expensive and differ based on the arithmetic formulae used to summarize statistics based on 4-local

neighborhood, 8-local neighborhood and global. Some of the models employ a complex expression while others focus on the entire application based on entropy and mathematical regression-based approaches are also formulated.

## C. Pre-Processing

These models are generally three-level breakdown hierarchy that offers enhanced assessment with an array of scaling approaches from eyeball to iris to white estimation as a function with a structure. Thus, making it a viable option for design and development of the texture model, that could effectively co-ordinate with each level of the eye independently along with factors associated with edges or region limitation. The statistical function can be considered as level-sensitive product that factors of every attributes and constraints. Proposed model could be applied for several biometrics that are classified into more than one level based on the attributes of biometrics in consideration. This model co-ordinates with each modular subsystem and redefined the requirement in dynamic manner which assessing the requirement simultaneously transforming the application based on enhancements.

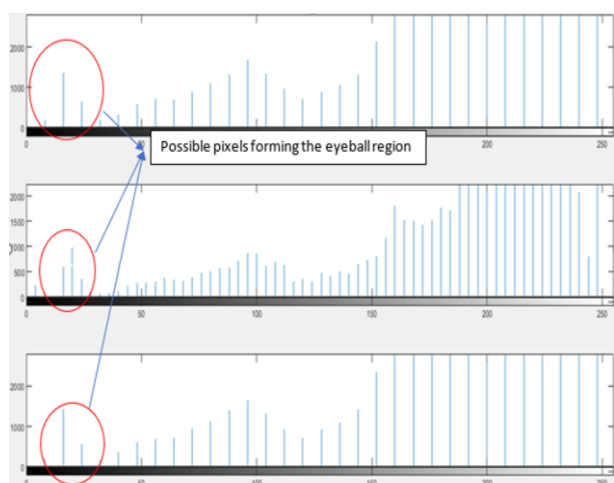
**Filtering/Smoothing Process:** The design and development of the filtering/smoothing model is based on the integration of the attributes of the 3x3 window as a collection of the several approaches such as 4-local neighborhood and 8-local neighborhood which supports the entropy and texture detection. In general, the filtering mask offers minimal changes in the pixel intensity values and then fends of the changes to smooth the image features which are sharp contrasts. It is quite important to exploit window size that offer smoothness to different level of the iris image, but on a similar note it preserves the features. The key motivation is always to provide a reasonable window size and contributing factors affect development and improvement of an iris image i.e. from initiation to refining the image regions. These models are generally three-level breakdown hierarchy that offers enhancing the assessment which array of scaling approaches from micro to macro estimation as a function with a structure. The prime function can be considered as level-sensitive product that factors of every attribute.

1. Eye-ball Level: This level is single-valued estimator that computes eyeball region assessment of eye projects as function of static factors such as pixel regions with fixed level threshold with minimum complex requirements from the iris image information.
2. IRIS Level: This level compute the effort required for completion of pixel region based on various factors such as diverse distribution of pixel information with or without prior analysis of IRIS images, the pixel region of iris is middle level of histogram analysis and complex requirements from the client information.
3. Pupil Level: This model assesses the eye information based on intermediate factors independently such as impact of diverse distribution of pixel information with or without prior analysis of IRIS images, the pixel region of iris is middle level of histogram analysis and complex requirements from the client information.

The model incorporates three different values for the fixed constants represented by the soft complexity metrics/factor with reference to the above levels. The key motivation for these set of models is always to provide the enough image analysis and with a detail understanding of the factors have the complexity metric.

The key motivation for these set of equations essential for assessment is based on the factors that contribute to the nominal pixel information that depends on the complexity metric of the image sub-block analysis in consideration. Further, the fixed analysis depends on the complexity metrics associated within the iris image and/or application.

The range and various in these sub-blocks of the eye image is evident that the model employs parameterized equation and exploits factors and attributes within the permissible constraints.



**Fig 1: Right (Image-2) eye original images with corresponding histograms of each layer with eyeball threshold**

**Segmentation Process:** The prime definition of segmentation process does not alter but the parametric set associated with the expression would be varying based on the factors, attributed and in accordance with the pixel distribution of IRIS image in consideration.

There are several approaches proposed that could be employed to identify border pixels that could investigated to exploit as edge pixels for different clusters of the iris images in consideration primarily 3-levels [21-26].

It is well-known fact that diagonal edges are less structured and are more forgiving regarding revealing borders than vertical or horizontal edges, are used as regions for segmentation process. We modify this approach based of the neighborhood direction to determine the variation measure of the pixel [10].

Let  $x_{mn}$  be the pixel under consideration at location  $(m, n)$  in an image of size  $M \times N$ .  $m=1, 2, \dots, M$ ; and  $n=1, 2, \dots, N$ . The variation intensity  $\hat{x}_{mn}$  may be defined as

$$\hat{x}_{mn} = \left( \sqrt{[W_1 * V^2] + [W_2 * H^2] + [W_3 * D^2]} \right) / (W_1 + W_2 + W_3).$$

$\{W_1, W_2, W_3\}$  are the weights associated with vertical, horizontal and diagonal variations of the pixel and for simulation purpose equal weights for three variations are used. The prime advantages of process are that it is simple to

use and has capability to develop reliable applications for managing complex and precise systems.

**Normalized Process:** Despite several approaches existing for normalization, there is no perfect and dependable approach that could be called good technique. An approach can be expected to balance the need of description or comprehend the image regions, but it is quite difficult to determine the size for new, modified pixels and clusters automatically.

Until it is fully defined, estimating the matching code is a difficult task due to various factors and attributes. The class of approaches differs mostly based on the mathematical sophistication, but they are directly related to size of the iris area available.

While some these models employ a complex expression based of differential equations and others mathematical regression. It is general understanding that the scale factors that are broadly classified based on three attributes that are namely: 1. IRIS Region, 2. Opening of eye (minimum 70%) & 3. Process Maturity.

The model could be exploited to a maximum extent to assess the effort needed for the completion of the entire project that consists of several heterogeneous modules. The size and complexity ratings could difference from one module to another. The prime means for utilizing the texture coding modules would ensure that the time required for the development of the system would reduce accordingly

#### D. Selection of Threshold Divers

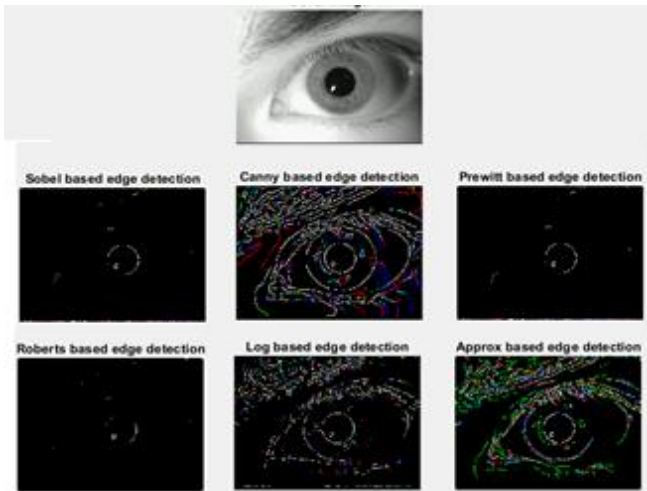
The thresholding scales of the iris are alter based on the class of the pixel distribution in consideration. At any given instance, this factor is combination of scale factors of that could precisely estimate histogram and order statistics to a convergent assessment.

The images are processed by the median filtering algorithm of size  $7 \times 7$  to determine the convergence of different centroid of each group i.e. eyeball, iris and pupil, and corresponding formations results would be displayed. Based on the classification and marking of each threshold could be utilized in addressing the issues that occur at each Region of Interest (ROI). The gradient lines are vital features to get measure of other characteristics associated with each ROI (in this case we limit our focus to IRIS region).

#### IV. IRIS SIGNATURE ESTIMATION

It is evident that there are various ecological reasons for unsuccessful estimating iris information but based on previous chapters size can be calculated from inherent features of GLT and GHT, the information that could be collected is another point.

Segmentation and edge detection tools are used to frame to perform the tasks in detection of changes in the images that influence the biometric based tools as it improves various approaches that use varied features in recognition and authentication-based development processes.



**Fig 2. Original Image with corresponding masks of the different edge detection operators**

The focus of segmentation methods is mostly focussed in the generation that is factored or analysed by changes in pixel intensity within or around the region, detection in directional changes in pixels that are incorporated to determine the basic image feature that interested are of pivotal i.e. isolated points, edges, and lines. It is a well-known fact that pixel intensity changes are often analysed by employing derivatives over the sub-blocks of image primarily 1<sup>st</sup> and 2<sup>nd</sup> order derivatives are of considerable interest in the field of image processing that is well suited for this purpose. The primary aim was to enhance the edge detection within the images so that they could acknowledge the concern benefits which in turn reduce segmentation closure. Edge detection, thresholding and segmentation would be used to ensure the iris image features for the unsupervised segmentation as presented in the figure 2

## A. Image Enhancement

As the fact that we need only the histogram of the image to segment it, segmenting images with Threshold Technique does not involve the spatial information of the images. Therefore, some problem may be caused by noise, blurred edges, or outlier in the image. That is why we say this method is the simplest concept to segment images. When the intensity distributions of objects and background pixels are sufficiently distinct, it is possible to use a single (global) threshold applicable over the entire image.

**Clustering Approach:** The frameworks were introduced for iris localization and estimation for different applications; unfortunately, their quality of the estimation was reducing i.e. their significant increase in the deficiencies, errors. With significant rise in the demand of the quality in iris region, clustering is becoming a popular method to address such issues based on the training datasets. For every cluster that could be reviewed by the expert over any of the clustering algorithm could be marked for all the modules in that group. In this phase, we employed the dataset that consists of clusters around 3 with different characteristics and functionalities as per iris images. These clusters are processed by the proposed algorithm to determine the convergence of centroid of each cluster or object, and corresponding formations results.

The determined centroid would have number of clusters

that are calculated according to various inbuilt factors such as the error ratio, exceptions, lines of code, faults are divided into each centroid. Based on the classification and marking of each cluster could be utilized in addressing the issues that occur at each centroid. The corresponding centroids of each cluster are formed based on similarities and relevant features that are divided into each centroid. Our proposed algorithm maintains the relation between each centroid based on all factors such as the eye ball, iris, pupil location, and other.

## B. Wavelet Decomposition

Wavelet decomposition plays a major role in development of images analysis which consumes half of the time of any transform while offering benefits of both spatial and transform domain. Based on the wavelet decomposition and the image features templates are captured whose quality is calculated. The advantages of spectral transforms such as transformation of a complex problem into easier problem and provides insight information for analysis and design. In addition, these localized features as separated sub-bands which not only improve the robustness but also reduce the complexity of the algorithms. The ability to analyze the edges structures in an image shows that sub-band decomposition would play a vital role in digital secured communication.

## C. Template Generation

This section will introduce some necessary definition needed for the derivation of the sensitivity measure and investigation of some measurement properties essential template

generation. Masks: Let  $M_{R,C} = \begin{bmatrix} W & L & \bar{W} \\ M & O & M \\ W & L & W \end{bmatrix}$  be any mask at R,

C pixel location. We denote the number of adjacent pixels surrounding a center pixel at the pixel location R, C. The binary pixels compared are  $|P_{R,C} - P_{R-1,C}| < 1$ , defined by the mask used and block size

**Coefficient Sensitivity Measure:** Let I be an image of size  $N_1 \times N_2$  and let  $I_B$  be the wavelet decomposition of an 8-bit color layer. Each of the decomposed layer must be divided into blocks, sub-blocks of 8 by 8, of  $M_1 \times M_2$  from the image  $I_B$ . Then the co-efficient templates are defined by

$$\gamma_i(m,n) = \frac{\beta_i}{\beta_{\max}}, \quad i = 0,1,\dots,7$$

where m and n are the block locations being analyzed and i defines the layer. General Algorithm for Calculation of template are

**Input** Input a wavelet coefficient of the Iris image.

Step 1: Divide the image into its 8 by 8 blocks with needed padding.

Step 2: Perform Haar wavelet decomposition.

Step 3: Determine the sensitivity measure to use.

Step 4: Divide each coefficient based on the sensitivity measure.

Step 5: Determine the mask size to use (default we are using 128).

Step 6: Calculate the value of  $\gamma$  for each block.  
 Step 7: Calculate the initial value for the threshold of  $\Gamma$ .  
 Step 8: Determine if  $\gamma$  meets the threshold and categorized  $\gamma$  into template mask.

**Output** template of iris.

The region within an image and arranges it (using special sorting scheme) in such an order to provide a means to randomly select coefficients by the closest one to measure. This offers the advantage that data can be analyzed and registered within the index, the palette or both therefore adding an extra level of security and assurance.

**Matching Score:** The region within an image and arranges it (using special sorting scheme) in such an order to provide a means to randomly select coefficients by the closest one to measure.

This offers the advantage that data can be analyzed and registered within the index, the palette or both therefore adding an extra level of security and assurance. The templates are compared using Euclidian Distance Measure and the corresponding score is recorded which acts in verifying the biometric templates.

### V. PROPOSED IRIS BIOMETRIC SYSTEM

In algorithm development, algorithm testing plays a vital role. Process contains the study of modeling, coding and maintenance of algorithm under different working conditions. Other way around it is the applying systematic disciplined and quantifiable method in the total product cycle of algorithm development and an engineering discipline that is concentrated phases in terms of biometric development. Formation of image capturing system, image processing toolboxes and innovation are playing a main role in biometric system.

Improved texture analysis; understanding thresholding and wavelets productivity; managing pixel enhancement requirements; and gathering user requirements comprises in analysis of image processing to implement iris analysis in several biometric systems.

There are two key points i.e., texture of the eye and how much of the localization factors can be presented within a defined region of image, and there are various ecological reasons for exploiting these estimating estimates.

#### A. Proposed Texture based IRIS Approach

**Input:** IRIS IMAGE: Any digital IRIS image that could be represented by a combination of binary signals could be employed as an input.

**Note:** We have employed several images but results are presented for 5-set images with both left and right eyes

**System 1:** Variation Calculation: This is a vital preprocessing block of the proposed framework that improves the localization of pixels to three levels as per prior discussion i.e. texture, edge and smooth while maintaining the order statistics of the data.

The Variation calculation was employed over each layer of the image. It is evident from the figure 3 the window size of 24 offer better results than the window size 8 but during the simulation we employ window size 8.

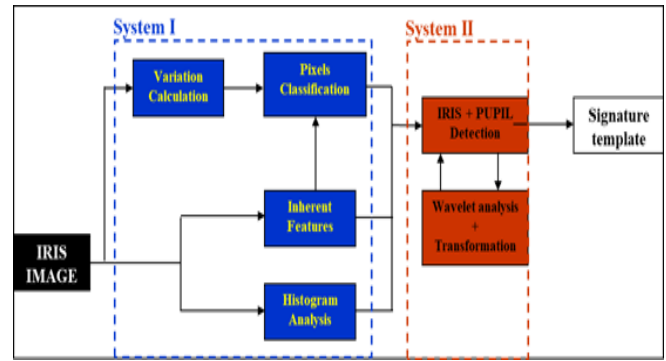


Fig 3. Block diagram of proposed texture-based IRIS approach

**Histogram Analysis:** In addition, purpose of histogram analysis is an effort to determine thresholds. Natural digital images generally produce a smooth histogram, characterized by gradual fluctuations throughout the entire intensity spectrum. The goal of histogram analysis is to minimize this affect overrun of edges from one region to another and maintain the natural appearance of the sub-image i.e. iris, pupil. We create a vector that records all every combination of the forward filter combination with pixel modifications.

**Inherent Features:** The determination of the variation measure pertaining to each pixel enables the algorithm to select at which locations would offer best information with reference to eyeball, IRIS and other information. If the global threshold measures (i.e. GHT & GLT) exceed a certain threshold level, the pixel is deemed as a suitable candidate for the IRIS (range of GHT), Eyeball (range of GLT) and/or other regions.

**Pixel Calculation:** The intensity measure is incorporated in selection of the pixels that are employed for smooth, texture and edge information bits within each cluster of the ROI i.e. eyeball, IRIS and Pupil. The pixels with higher measure are given high priority to be boundary based edge pixels while the pixels with a low intensity measure are part of cluster. In addition, the pixels with a higher priority could be employed for edge information bits.

$$\begin{aligned} \text{if } x_{mn} > \alpha_2 & \quad \text{edge information bits} \\ \text{if } \alpha_2 > x_{mn} > \alpha_1 & \quad \text{ROI information bits} \\ \text{if } \alpha_1 > x_{mn} & \quad \text{smooth information bits} \end{aligned}$$

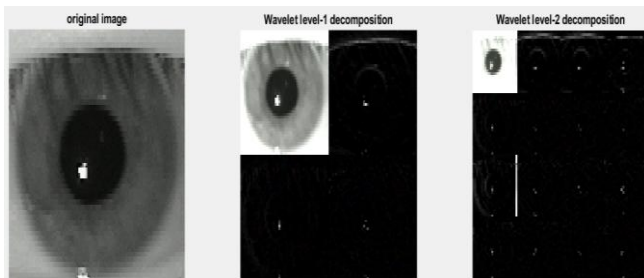
Data cannot be analyzed into the white or black pixels for smooth area as it could introduce no valuable analysis. This motivated us in applying wavelet and analyzing the wavelet coefficients in each sub-image.

#### B. System 2:

We firmly believe that need of a image effective detection and recognition method with un-supervised estimation of ROI associated with eye images. We introduce several existing ubiquitous models (supervised and un-supervised) of performance is benchmarked for analyzing the effectiveness of proposed algorithm. The proposed model is hybrid function that incorporates inherent features and global threshold properties into a parametric model with tuning parameters as and when required.

**IRIS and EYEBALL Detection:** It is well known fact that the IRIS and EYEBALL are concurrent to one another that could be better estimated by identifying the eyeball region and grow around to reach the iris region of the image. The reliable detection of IRIS and eyeball has become an important and essential component that is critical in iris recognition based biometric system. They are several techniques and methodologies exploited to design and develop an hybrid method that detects models in terms of pixel variation detection, with the vast majority of thresholding to offer the necessary support. These attributes also provide a means to understanding the feasibility and planning of with a corresponding monitoring and control at each phase.

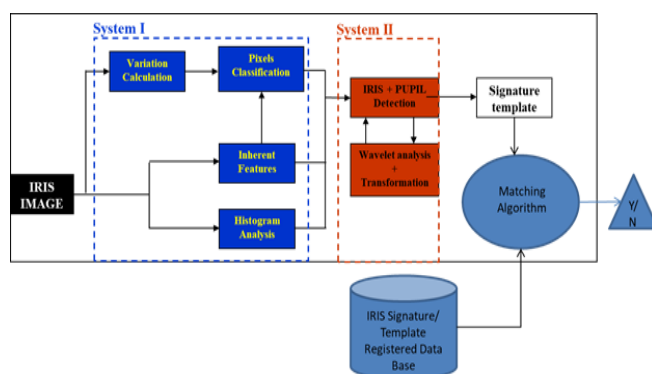
**Wavelet:** It is often found that the significant portion of the information alters with the change in luminance and angle with which the image has been captured by acquisition system for the registration of the eye image and process followed to collect the ROI of IRIS image. Henceforth, we exploit the wavelet transform to transform the ROI of IRIS image to the wavelet packets that could be helpful in identifying the wavelet coefficients for capturing the template information associated with IRIS image.



**Fig 4. IRIS extracted from original image with corresponding wavelet decomposition for Level-1 and Level-2**

Figure 5 presents a detail illustration of the iris region extracted from the eye information based on the proposed system and the corresponding wavelet decomposition of level-1 and level-2. It is also evident that the template deduced from the wavelet decomposition could be robust to various affine transformations and offer the features based on spatial and frequency domain. We exploit the wavelet transforms that are limited to level-2 for the proposed system that ensure the coefficients are retrieved in a structured manner to design the template.

**Outputs Signature Template:** The stream of digital data with unique information regarding IRIS



**Fig 5. Block diagram of proposed matching based IRIS approach**

## VI. RESULT ANALYSIS

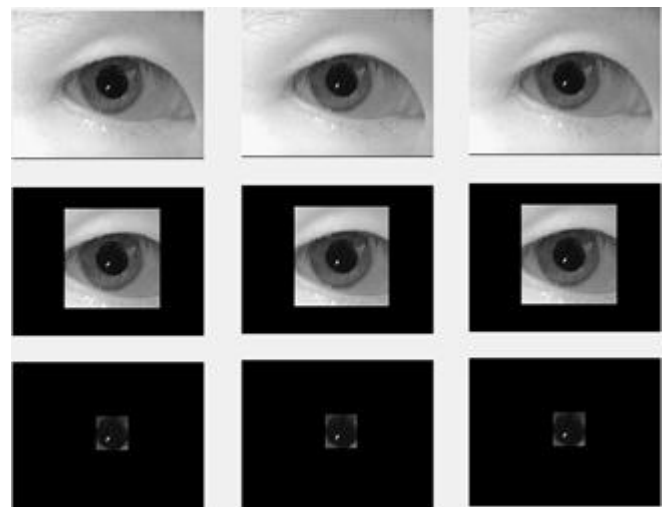
When compared to traditional measures the present approach shows a better output. Hence based on optimization techniques automated test generation techniques are to be developed to lessen the calculation of feature set and accuracy of testing. So many optimization methods are used on these iris biometric test-data and their applicable characteristics, when applied to these situations stay relatively unknown. For a given eye information as an input image and getting quality signature output based on IRIS information from the proposed approach under different working scenarios along with corresponding pro and cons is called as testing.

**Table-I: Threshold Analysis**

Images	9x9	7x7	5x5	Lower Threshold	Higher Threshold
IRIS-L-1	62	60	55	35	105
IRIS-R-1	62	58	57	33	103
IRIS-L-2	63	62	60	37	107
IRIS-R-2	62	57	56	32	102
IRIS-L-3	63	63	58	38	108
IRIS-R-3	60	60	61	35	105
IRIS-L-4	66	65	59	40	110
IRIS-R-4	62	59	52	34	104
IRIS-L-5	62	62	55	37	107
IRIS-R-5	63	60	58	35	105

### A. Threshold Analysis

Thus, from the table 1 (we have chosen a set of images for presentation) it is evident we taken minimum of all lower thresholds and maximum of all higher thresholds. In this thesis, for the database taken that contain both left and right eye. We conduct a detail simulation-based study based on the actual bench-marked information pertaining to the IRIS data sets for understanding the impact of Global Lower Threshold (GLT) and Global Higher Threshold (GHT). In addition, an application histogram-based enhancement for iris region is exploited examining various factors associated with the IRIS framework of data representations. The data sets illustrated based on the various measures using threshold analysis.



**Fig 6. The original images and corresponding tentative IRIS ROI after GHT and GLT for Right Eye Image-6**

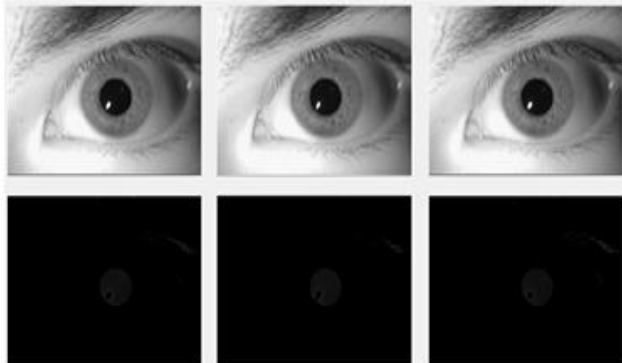


It is evident from the above figure 6, global threshold-based assessment approach offers significantly improved results for the selection of iris ROI in consideration in comparison with other commonly employed threshold approaches.

**B. Cluster Analysis**

**Table-II: Cluster Analysis**

Eye ball	iris	pupil	other
0-48	60-120	150-225	Eye ball reflection, eye brows
12-52	66-112	155-220	Eye ball reflection, eye brows
23-50	70-110	160-220	Eye ball reflection, eye brows



**Fig 7. Original Image with corresponding eyeball based on thresholding GLT**

It is evident from table 2, the effectiveness of the cluster analysis. We handle thresholding and segmentation demonstrate viably. Texture estimation process in created hybridization apparatus result as takes after for advancement issues with respect to information proficiency. By using eyeball region estimation performance or benchmarks rate is calculated as presented in the figure 7.

**C. Matching Score**

Initial investigation to test robustness, was based on the randomness of the different iris images from the eye image in consideration as illustrated in table 3. Hence, to measure the robustness of the proposed system, we need to measure the distance between the corresponding templates in the image.

$$Dis_{val} = \text{dist}(O, V) \text{ and } Dis_{ver} = \text{dist}(O, N)$$

Where, “Dis<sub>val</sub>” is defined as the distance between the original template ‘O’ and its corresponding validation template ‘V’. “Dis<sub>ver</sub>” is defined as the distance between the original ‘O’ and the corresponding new verification template ‘N’ iris image.

**Table-III: Distance metrics based on proposed measure**

Images	Euclidian	Euclidian	Proposed	
	Fixed	Random	Normal	Weighted
IRIS-L-1	81.38	75.47	79.22	95.72
IRIS-R-1	80.43	70.94	75.55	94.54
IRIS-L-2	73.93	69.91	73.12	90.03
IRIS-R-2	81.60	75.13	83.18	84.19
IRIS-L-3	87.72	79.84	84.62	92.18
IRIS-R-3	78.53	75.97	83.97	91.57
IRIS-L-4	83.08	84.61	85.02	89.22
IRIS-R-4	77.33	75.51	79.52	85.77
IRIS-L-5	84.93	84.72	78.98	88.31
IRIS-R-5	85.11	88.09	84.56	89.95

**D. Robustness Testing**

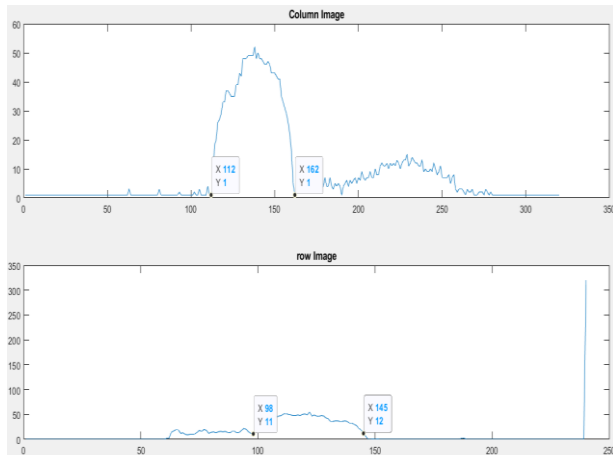
**Table- IV: Threshold metrics associated GLT based on Histogram Analysis for three layers of IRIS test images**

Images	Global Lower Threshold		
	Lower	Higher	Peak
IRIS-L-1	1,9,1	41,37,41	17,17,17
IRIS-R-1	1,9,1	41,33,41	17,17,17
IRIS-L-2	25,9,25	0,37,0	49,21,49
IRIS-R-2	1,9,1	41,37,41	17,21,17
IRIS-L-3	1,9,1	41,33,41	17,17,17
IRIS-R-3	1,5,1	41,41,41	17,17,17
IRIS-L-4	1,5,17	41,33,0	17,17,49
IRIS-R-4	1,5,1	33,29,33	17,17,17
IRIS-L-5	1,9,1	41,37,41	17,21,17
IRIS-R-5	25,9,25	0,41,0	49,25,49

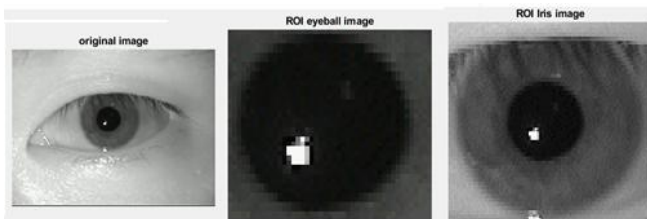
It is also called noise analysis, orientation analysis or module testing as it tests all the internal control structures and working process of a program, as compared to the properties shown to the end-user. An internal process structure of the iris biometric system and image processing skills are applied to design and develop test cases in this method we provide the input data i.e. eye image data and get output i.e. iris information. Robust testing process is applied at the unit level, integration level and system levels of the process of algorithm analysis. The validness of proposed system depends precisely on the eyeball extraction which further helps in identification of possible ROI of the IRIS.

**Table- V: Threshold metrics associated GHT based on Histogram Analysis for three layers of IRIS test images**

Images	Global Higher Threshold		
	Lower	Higher	Peak
IRIS-L-1	51,59,51	99,107,107	89,93,89
IRIS-R-1	51,59,51	91,87,91	73,81,73
IRIS-L-2	75,59,75	107,107,107	97,101,97
IRIS-R-2	51,59,51	107,103,107	97,97,97
IRIS-L-3	51,59,51	147,151,147	145,145,145
IRIS-R-3	51,55,51	171,159,163	145,149,145
IRIS-L-4	51,55,67	107,115,107	97,109,97
IRIS-R-4	51,55,51	147,159,147	145,149,145
IRIS-L-5	51,59,51	107,107,107	97,101,97
IRIS-R-5	75,59,75	123,119,123	97,101,97



**Fig 8. Possible ROI location within the iris image for eyeball**



**Fig 9. Original image with ROI eyeball image along with extracted IRIS image using proposed system**

It is evident from figure 8 & 9 that the proposed algorithm identifies the ROI of eyeball and grows in square block size around ROI of eyeball with an intention to cover IRIS ROI completely which is attained by the proposed with high reliability based on the global thresholds i.e. GLT and GHT from features of the original IRIS image as expressed in the tables 4 and 5 for GLT and GHT respectively.

The validness of proposed system depends precisely on the quality of the eyeball extraction which further helps in identification of possible ROI of the IRIS. The Table. 6 illustrates different the test images that are widely examined and the variability with reference to the eyeball and the closure of the iris area are studied in detailed and analyzed. This analysis produces a full look of robust of the proposed algorithm in terms of eyeball detection as that pave's ways for the completed iris recognition and detection.

**Table- VI: Robustness of the proposed approach in detection of eyeball in different luminance and captured of IRIS test images**

Images	EYE BALL INFORMATION		
	X-Coordinates	Y-Coordinates	Image
IRIS-L-1	98, 145	112, 161	
IRIS-R-1	78, 131	128, 178	

IRIS-L-2	86, 130	144, 190	
IRIS-R-2	91, 146	164, 218	
IRIS-L-3	88, 128	132, 176	
IRIS-R-3	106, 158	147, 196	
IRIS-L-4	104, 157	135, 206	
IRIS-R-4	104, 160	149, 210	
IRIS-L-5	101, 159	154, 209	
IRIS-R-5	85, 146	165, 218	
IRIS-L-6	72:125	145:197	
IRIS-R-6	87:144	140:195	

### E. Error Testing

It views the biometric algorithm as an error testing functionality with no knowledge of internal details of template generation or its implementation". (Having no information of inner details of the template the proposed algorithm creates it by a fusion of fixed and random parameters based on the wavelet transforms). The validation of the template is viewed as error testing which need to be addressed by the following types of the errors which could be better understood confusion matrix associated with false acceptance rate (FAR) and False Rejection Rate (FRR).

The possible errors are defined when false result could be detected true which in turn classify the errors into two types i.e.

- Error in authenticating a person as a registered person but has not registered in data base
- Error in authenticating a person as an un-registered person but has registered in data base

There is no necessity to know the implementation knowledge in this process which is an added benefit. But the performance of the verification and authentication system depends on ensuring that false positives are minimized. The proposed system could offer a desired acceptance ratio in terms of false positive and true negatives if the percentage of iris information is greater than 55% minimum. With increase in the iris information the acceptance ratio would increase drastically as presented in the figure 10.

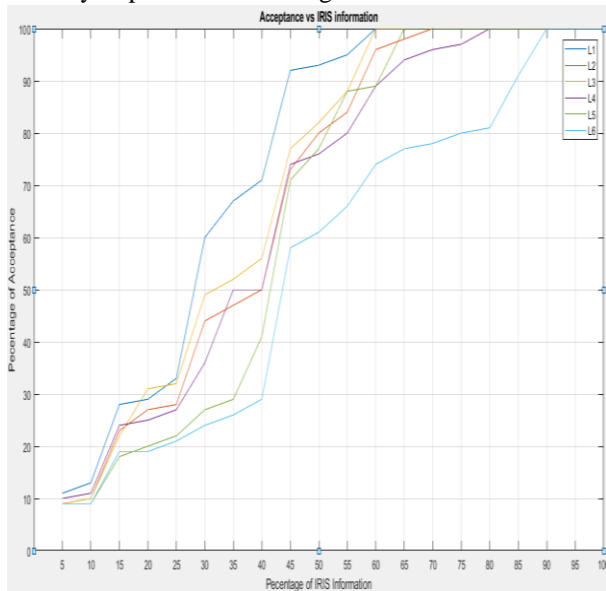


Fig 10: Percentage of IRIS information vs Percentage of Acceptance

### VII. CONCLUSION

With ever increasing demand for a high accuracy, consistent and efficient biometric approach that could register, analyze and validate along with verification of IRIS image, several algorithms came into existence. Unfortunately, several factors were proposed which could have a considerable effect on proposed IRIS based biometric estimation model in such as:

- 1 There are considerable number of correlated factors such as signature of iris, length of the signature, regional features of iris, quality of iris, system attributes and image processing components that influence the design and development process.
- 2 There is constant evolving of new image processing operations for design of texture and localized iris based applications; hence in this dissertation we employ parametric objective function for localization.
- 3 The complexity metrics of iris signature in consideration can only be approximated but not measured we employ matching model for better comparison.

Based on the detail analysis, it was evident that better estimates from the proposed texture based estimators can be obtained by focusing on relation between the localized features and type of image processing attributes and corresponding matching impact and complexity of the system. It is evident that for any texture based model, the training of the model on a wider scale on wavelet domain collected is essential. Henceforth, a standard and benchmarked repository (CASIA IRIS DATASET) is used to verify and compare the proposed approach. The proposed approach considers the

historical data and new weighted statistical based objective function for robust, consistent, effective and efficient texture based iris recognition via wavelet estimation model as proved in simulation results. With constant evolving of new development tools for different complex applications and paradigms would require an adaptive image processing estimator that can offer accurate and reliable segmentation at low contrast which would be greatest challenge in future. A conclusion section is not required. Although a conclusion may review the main points of the paper, do not replicate the abstract as the conclusion. A conclusion might elaborate on the importance of the work or suggest applications and extensions.

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