

## Efficient Method for Color Iris Localization

Esraa G. Daway ${ }^{1 *}$, Loay E.George ${ }^{2}$, Radhi Sh.Hamoudi ${ }^{1}$

${ }^{1}$ Department of Physics, College of Education, Almustanserya University, Baghdad, Iraq ${ }^{2}$ Department of Computer Science, College of Science, University of Baghdad, Baghdad, Iraq


#### Abstract

Iris detection is considered as challenging image processing task. In this study efficient method was suggested to detect iris and recognition it. This method depending on seed filling algorithm and circular area detection, where the color image converted to gray image, and then the gray image is converted to binary image. The seed filling is applied of the binary image and the position of detected object binary region (ROI) is localized in term of it is center coordinates are radii (i.e., the inner and out radius). To find the localization efficiency of suggested method has been used the coefficient of variation (CV) for radius iris for evaluation. The test results indicated that is suggested method is good for the iris detection.


Keywords: iris, localization, Isolation, recognition.

## الطريقة (الفعاليه لتحديد القزحيه الملونـه

اسراء كاطع دواي 1**، لؤي ادور جورج²، راضي شدهان الطويل1
1فسم الفيزياء ، كلية النزبيه ، الجامعه المستتصريه، بغداد، العراق
2قفم الحاسبات، كلية العلوم، جامعة بغداد، بغداد، العراق


## 1. Introduction

With the increasing emphasis on security, automated personal identification based on biometrics has been receiving extensive attention over the past decade. Biometrics aims to accurately identify each individual using different physiological or behavioral characteristics, such as fingerprints, iris, face, retina, gait, palm-prints and hand geometry (Jain et al., 1999; Zhang,2000). Recently, iris recognition is becoming an active topic in biometrics due to its high reliability for personal identification [1] (Jain et al., 1999; Mansfield et al. 2001; Daugman, 2003). Iris localization is a key step in iris recognition; it isolates the iris part in eye image by detecting both the inner and outer boundaries of iris area. The overall performance of an iris recognition system is highly related to the localization accuracy [2]. The accurate pupil detection, as the inner iris boundary, is the most important and the first step in iris localization (Kheirolahy et al., 2009). Numerous methods were proposed to locate the pupil; they based on the fact that the pupil edge can be estimated as circular shape.

[^0]
## 2. The Iris

Human eye iris is a visible color ring bounded by the pupil (the dark opening) and white sclera, as depicted in Figure-1.


Figure 1- Anterior surface of the iris
The iris is suspended in aqueous humor, behind the cornea but in front of the lens as shown in Figure2.


Figure 2- Anterior segment of the eye (Modified from [3]).
The iris divides the space between the lens and the cornea into the anterior chamber (between the cornea and the iris) and posterior chambers (between the iris and the lens). The periphery of the iris is attached to its root at the iridocorneal angle of the anterior chamber where it merges with the tissue inside the eye: the ciliary body and trabecular meshwork. The free edge of the iris is known as the pupillary margin or pupillary ruff (the boundary of the pupil). The formation of the human iris begins during the third month of gestation; its distinguishable pattern is completed by the eight month of gestation, but pigmentation continues into the first year after birth [4]. The size of the iris varies from person to person with a range of 10.2 to 13.0 mm in diameter [5], an average size of 12 mm in diameter, and a circumference of $37 \mathrm{~mm}[6,7]$.
3. The Proposed Iris Localization System

In this work an iris localization system is developed and implemented. It consists of three main stages, as shown in Figure-3.


Figure 3-The main stages of the Iris Localization system
In the following the steps of theses main stages are given:

### 3.1 Load Image Data

As first stage the color image data is located from the image data as raster file consist of 3 colors component (Red, Green, Blue), and then finding the lightness according to the following equation (1), as shown in Figure-4.

$$
\begin{equation*}
g(x, y)=\frac{1}{3}(\operatorname{Red}(x, y)+\operatorname{Green}(x, y)+\operatorname{Blue}(x, y)) \tag{1}
\end{equation*}
$$



Figure 4- An example of fundus area

### 3.2 Isolation for ROI (Region of interest)

In put: color image with band R (wid,hgt), G (wid,hgt), B (wid,hgt)
Output: $\mathrm{rad} / / \mathrm{Rad}$ in for ROI
This step includes sub steps as following:
A. Calculate the Histogram for the gray image using the equation:
$\operatorname{Hist}(i)=\frac{N(i)}{N}$
Where, $N=$ Width $_{\text {Image }} x$ Height $_{\text {Image }}$ is the size of image. $N_{i}$ is the frequency of occurrence of gray level (i).
B. Binarization by Thresholding: if implies the following
(i) Finding Threshold value G

$$
G=0
$$

While $\operatorname{His}(G)>\operatorname{His}(G+1) \operatorname{And} \operatorname{His}(G)>100$

$$
G=G+1
$$

End while
(ii) Make binarization by thresholding. The central part holds the darkest pixels (which are flagged 1, while other pixels are flagged 0 ).
Set the array dimension: $\operatorname{Bin}(\mathrm{Wm}, \mathrm{Hm})$
For $Y=0$ To Hm: For $X=0$ To Wm
If $\operatorname{Gry}(X, Y)<=$ Threshold Then $\operatorname{Bin}(X, Y)=0$ Else $\operatorname{Bin}(X, Y)=1$
Next X: Next $Y$
C. Remove the small isolated areas (i.e., islands/gaps) using seed filling algorithm.
D. Calculate the Center Point Coordinates and Radius of the Bright Area, using the following equation:
$x_{c e n}=\frac{1}{n} \sum_{i=1}^{n} x_{i}, \quad s_{y}=\frac{1}{n} \sum_{i=1}^{n} y_{i}, \quad R=\sqrt{\frac{n}{\pi}}$
Where, $n$ is the number of collected white pixels; $\left(x_{i}, y_{i}\right)$ are the coordinates of the $\mathrm{i}^{\text {th }}$ collected bright pixel.
E. Establish the Mask of the Brightest Area, Bin(), that represents the Isolated ROI region area, as shown in Figure-5. The following shown steps are for establishing the mask:
For all y value

$$
\begin{aligned}
& d y=y-y_{c} \\
& \text { If dy } d y \text { then } \\
& \text { For all } x \text { value } \\
& \quad d x=x-x_{c} \\
& \quad \text { if }(d x)^{2}+(d y)^{2}<R \text { then } \operatorname{Bin}(x, y)=1 \text { else } \operatorname{Bin}(x, y)=0 \\
& \text { Next }
\end{aligned}
$$

Figure-6 shows a sample of the established mask:


Figure 5- The isolation of ROI as circle region


Figure 6- The Mask Image output of do Isolation for ROI (Region of interest)

### 3.3 Iris Isolation

In put: G (wid,hgt), B (wid,hgt) with size $\mathrm{x}, \mathrm{y}$
Output: Iris Isolation.
In this stage the iris is isolated from the ROI region by using following steps:
A. Calculate the Mean ( $m$ ) \& Standard Deviation ( $\sigma$ ) [8] of the Pixels belong to ROI by applying equations (4) and (5).
For all x and y values DO
If $\operatorname{Bin}(x, y)=1$
$m=\frac{1}{N} \sum \boldsymbol{G}(\boldsymbol{x}, \boldsymbol{y})$
$\sigma=\sqrt{\sum_{x, y} G^{2}(x, y)-\boldsymbol{m}^{2}}$
B. Apply Linear Contrast Stretching and calculate the mean of the established stretched image, the applied Contrast Stretching for gray image was performed only on the pixels belong to ROI .The stretching step is applied using the following equation (6):
Ics $=\left\{\begin{array}{lr}0 & \text { if } G \leq \text { Min } \\ 255 \frac{(G-\text { Min })}{\text { Max Min }} & \text { if Min }<G<\text { Max } \\ 255 & \text { if } G \geq \text { Max }\end{array}\right.$
Where, Min $=m-\sigma$ and $\operatorname{Max}=m+\sigma$.
C. Do Binarization by Thresholding: The gray image is converted to binary image using the mean value as threshold value.
For all x value
For all y value
If $\operatorname{Bin}(X, Y)=1$ Then
If $\operatorname{tImg}(\mathrm{X}, \mathrm{Y})>=$ Mean $\operatorname{Then} \operatorname{tImg}(\mathrm{X}, \mathrm{Y})=1$ Else $\operatorname{tImg}(\mathrm{X}, \mathrm{Y})=0$
End if

## Next y

Next x
D. Do Tiling; Repeatedly, to remove the white points appeared in the binary image, after converting the image to binary, the image conform two regions as same circle (white and background black). The edge region includes some noise points, to remove this point we used tiling:
$\operatorname{tImg}(\mathrm{X}, \mathrm{Y})=\operatorname{Tiling}\left(\mathrm{I}_{\mathrm{cs}}(\mathrm{X}, \mathrm{Y})\right)$
E. Apply seed filling to remove gaps and islands in the iris region [9].
F. Remove the boundary points using a window (with size $3 \times 3$ ):

Set $\operatorname{Bin}(X, Y)=1$ iff $(x, y)$ is a location not close to boundary of the region
Otherwise set $\operatorname{Bin}(\mathrm{X}, \mathrm{Y})=0$
G. By using the modified mask, $\operatorname{Bin}()$, re-establish the binary image from the gray image.
H. Invert the Binary Representation of the binary image using the following:

If $\operatorname{Bin}(\mathrm{X}, \mathrm{Y})=1$ Then $\operatorname{tImg}(\mathrm{X}, \mathrm{Y})=1-\operatorname{tImg}(\mathrm{X}, \mathrm{Y})$
I. Determine the Center Coordinates ( $\mathrm{X}_{\text {cen }}, \mathrm{Y}_{\text {cen }}$ ) and Radius $\left(\mathrm{R}_{\mathrm{d}}\right)$ for the iris pixels (i.e., the pixels whose binary value is 1 ):
Set $\mathrm{S}_{\mathrm{x}}=0: \quad \mathrm{S}_{\mathrm{y}}=0: \mathrm{n}=0$
For all $\mathrm{X} \in[0$, ImgWid-1]
For all $\mathrm{Y} \in[0, \mathrm{ImgHgt}-1]$
If $\operatorname{Bin}(\mathrm{X}, \mathrm{Y})=1$ and $\operatorname{timg}(\mathrm{X}, \mathrm{Y})=1$ Then

$$
S x=S x+X: \quad S y=S y+Y: \quad n=n+1
$$

End if
Next Y
Next X
$X_{c e n}=\frac{S_{x}}{n}, \quad Y_{c e n}=\frac{S_{y}}{n}, \quad R_{d}=\sqrt{\frac{n}{\pi}}$
J. Apply Radial Scan to Collect the External Boundary points, and then determine the radius array, $\operatorname{Rad}()$.
K. Determine the mean ( $\mathrm{R}_{\text {mean }}$ ) and standard deviation ( $\sigma_{\text {radius }}$ ) of all Radii collected within all sectors.
L. Re-determine the center coordinates ( $\mathrm{X}_{\text {cen }}, \mathrm{Y}_{\text {cen }}$ ) and radius values ( $\mathrm{R}_{\mathrm{d}}$ ), to get more Accurate Values. The following criterion has been applied to filter out the abnormal radii elements values and then re-determine the parameters ( $\mathrm{X}_{\text {cen }}, \mathrm{Y}_{\text {cen }}, \mathrm{R}_{\mathrm{d}}$ ):

$$
\text { if }\left|\operatorname{Rad}(i)-R_{\text {mean }}\right|>3 \sigma_{\text {radius }} \text { then exclude the } i^{\text {th }} \text { radius element }
$$

M. The determined center coordinates and radius are used as the geometric parameters for defining the external radius of the iris. They used to establish iris area mask which is circular area as shown in Figure-7.


Figure 7- A Sample of Iris Isolated Area

## 4. Result and Discussion

A number of the iris images with different color and direction have been used In this study the standard dataset .The database contains $3 \times 128$ iris images (i.e. 3 simple for the left iris and 3 sample for the right iris for (64) Persons ).The images are: 24 bit - RGB, $576 \times 768$ pixels, file format: PNG. The irises were scanned by TOPCON TRC50IA optical device connected with SONY DXC-950P 3CCD camera [10]. Figure-8 illustrates the iris images (some data base images captured from left direction).


Figure 8- Some of the database images captured from left eye


Figure 8- Some of the database images captured from left eye (continue)
Table-1 presents the mean $(\mu)$, standard division $(\sigma)$ and the coefficient of variation (CV) for iris radius, from this table we can see the coefficient of variation $(\mathrm{CV})$ values are small due to the high accuracy of the iris distinguish.

Table 1- The mean $(\mu)$, standard division $(\sigma)$ and coefficient of variation $(\mathrm{CV})$ for determined iris radius

| No. | $\mu$ | $\sigma$ | C.V | No. | $\mu$ | $\sigma$ | C.V |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 245.366 | 4.063 | 0.0166 | 33 | 248.368 | 1.274 | 0.0051 |
| 2 | 243.913 | 1.316 | 0.0054 | 34 | 244.900 | 0.640 | 0.0026 |
| 3 | 244.601 | 7.140 | 0.0292 | 35 | 250.723 | 2.668 | 0.0106 |
| 4 | 249.300 | 2.848 | 0.0114 | 36 | 248.887 | 3.205 | 0.0129 |
| 5 | 246.774 | 3.062 | 0.0124 | 37 | 251.006 | 2.736 | 0.0109 |
| 6 | 244.507 | 2.887 | 0.0118 | 38 | 258.747 | 9.022 | 0.0349 |
| 7 | 254.618 | 1.246 | 0.0049 | 39 | 260.230 | 8.644 | 0.0332 |
| 8 | 238.187 | 4.879 | 0.0205 | 40 | 260.230 | 8.644 | 0.0142 |
| 9 | 238.669 | 3.778 | 0.0158 | 41 | 247.494 | 6.028 | 0.0244 |
| 10 | 246.709 | 3.147 | 0.0128 | 42 | 244.542 | 1.910 | 0.0078 |
| 11 | 246.826 | 2.226 | 0.0090 | 43 | 256.918 | 3.676 | 0.0143 |
| 12 | 250.206 | 1.543 | 0.0062 | 44 | 241.571 | 6.236 | 0.0258 |
| 13 | 251.216 | 5.188 | 0.0206 | 45 | 241.964 | 4.595 | 0.0190 |
| 14 | 253.342 | 3.458 | 0.0136 | 46 | 252.530 | 4.724 | 0.0187 |
| 15 | 250.129 | 1.341 | 0.0054 | 47 | 244.215 | 2.025 | 0.0083 |
| 16 | 250.615 | 6.202 | 0.0247 | 48 | 245.878 | 2.505 | 0.0102 |
| 17 | 249.464 | 1.931 | 0.0077 | 49 | 243.708 | 0.993 | 0.0041 |
| 18 | 246.205 | 3.627 | 0.0147 | 50 | 243.127 | 1.796 | 0.0074 |
| 19 | 246.936 | 17.125 | 0.0693 | 51 | 247.568 | 2.560 | 0.0103 |
| 20 | 249.239 | 8.309 | 0.0333 | 52 | 247.118 | 5.847 | 0.0237 |
| 21 | 239.916 | 4.128 | 0.0172 | 53 | 256.499 | 4.504 | 0.0176 |
| 22 | 252.797 | 1.151 | 0.0046 | 54 | 249.274 | 5.239 | 0.0210 |
| 23 | 245.984 | 1.211 | 0.0049 | 55 | 246.530 | 3.032 | 0.0123 |
| 24 | 250.573 | 1.257 | 0.0050 | 56 | 251.315 | 4.922 | 0.0196 |
| 25 | 238.714 | 9.779 | 0.0410 | 57 | 251.695 | 1.651 | 0.0066 |
| 26 | 249.314 | 2.452 | 0.0098 | 58 | 238.648 | 3.805 | 0.0159 |
| 27 | 253.718 | 1.576 | 0.0062 | 59 | 245.966 | 5.402 | 0.0220 |
| 28 | 243.645 | 4.908 | 0.0201 | 60 | 238.216 | 6.852 | 0.0288 |
| 29 | 246.618 | 3.864 | 0.0157 | 61 | 249.473 | 5.221 | 0.0209 |
| 30 | 241.315 | 4.190 | 0.0174 | 62 | 241.229 | 1.399 | 0.0058 |
| 31 | 253.175 | 7.484 | 0.0296 | 63 | 252.406 | 3.863 | 0.0153 |
| 32 | 248.888 | 1.775 | 0.0071 | 64 | 251.038 | 3.1469 | 0.0125 |

Figure-9 shows the histogram of the iris radii computed from the database images. The distribution is close to Gaussian (normal) distribution.


Figure 9- The histogram of the determined iris radius

## 5. Conclusion

In this study an efficient suggested method for iris detection for iris images. The results indicated the coefficient of variation (CV) normal distribution, and we can see the suggested algorithm has High accuracy in the iris detection.

## References

1. Li Ma, Tieniu Tan, and Fellow. 2004. Efficient Iris Recognition by Characterizing Key Local Variations. IEEE Transactions on Image Processing,13(6).
2. Jarjes, A.A., Kuanquan Wang, and Mohammed, G.J. 2010.Iris localization: Detecting accurate pupil contour and localizing limbus boundary. IEEE, 1:349-352.
3. Davis-Silberman, N., and Ashery-Padan, R. 2008. Iris Development Vertebrates Genetic and Molecular Considerations, Brain Research, 1192(5):1728.
4. Muron, A. and Pospisil, J. 2000. The Human Iris Structure and Its Usages. 39:87-95.
5. Caroline, P. J. and André, M. P. 2002. The Effect of Corneal Diameter on Soft Lens Fitting, Part1, Contact Lens Spectrum, 17(4): 56.
6. Somying Thainimit, Luís A. Alexandre, and Vasco M. N. de Almeida. 2013. Iris Surface Deformation and Normalization, IEEE, pp:501-506, 4-6.
7. Forrester J. V., Dick, A. D., McMenamin, P. G., and Lee, W. R. 2001. The Eye: Basic Sciences in Practice, W.B. Saunders.
8. Maurice D., Weir, and Joel Hass. 2010. Thomas Calculus Early Transcendentals, Twelfth Edition, Pearson Education.
9. Henrich, D.1993. Space-Efficient Region Filling in Raster Graphics. The Visual Computer: An International Journal of Computer Graphics.
10. Michal Dobeš and Libor Machala, Iris Database, website: http://www.inf.upol.cz/iris/.

[^0]:    *Email: esraado2007@yahoo.com

