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# Towards more accurate iris recognition system by using hybrid approach for feature extraction along with classifier

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## **ABSTRACT**

Iris recognition become one of the most accurate and reliable steadfast human biometric recognition system of the decad. This paper presents an accurate framework for iris recognition system using hybrid algorithm in preprocess and feature extraction section. The proposed model for iris recognition with significant feature extraction was divided into three main levels. First level is having pre-processing steps which are necessary for the desired tasks. Our model deploys on three types of datasets such as UBIRIS, CASIA, and MMU and gets optimal results for performing activity. At last, perform matching process with decision based classifier for iris recognition with acceptance or rejection rates. Experimental based results provide for analysis according to the false receipt rate and false refusal amount. In the third level, the error rate will be checked along with some statistical measures for final optimal results. Constructed on the outcome the planned method provided the most efficient effect as compared to the rest of the approach.

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#### 1. INTRODUCTION

In the current days, iris recognition as a functional feature of biometric is a chief biometric development. Human eye iris performances as substantial work in vast sympathy of a human being. The use of science and technology-based knowledge made an efficient effort in the field of biometrics authorization system where it used for individual can adapt it [1]. Biological science are used in different field for different purpose and the function of physiognomies are used for independence fingerprints, face and patterns, retina in iris system. Use of interactive physiognomies contemplate voice, handwritten autograph. One of the most reliable biometric is the iris, due to its stability, uniqueness and noninvasive nature [2]. Various extraction methods of iris features from iris image exist. The biometric authentication of iris image class schemes like local-global graph methodology, this is a graph based method for authentication of extracted features and many others for iris authentication need to be studied extensively. The existing schemes will be studied thoroughly and give experimental based results in order to have in-depth knowledge. This is because to have a state-owned of the art research for iris class recognition. Recognition in general and iris feature extraction is specific in security [2], [3].

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#### 2. RESEARCH METHOD

Methods for enhancing low contrast iris image class at preprocessing phase and significant features extraction process. This method for enhancing low contrast iris image class and significant features extraction through the proposed hybrid approach. For more information, the proposed technique working and explanation are present in Figure 1.

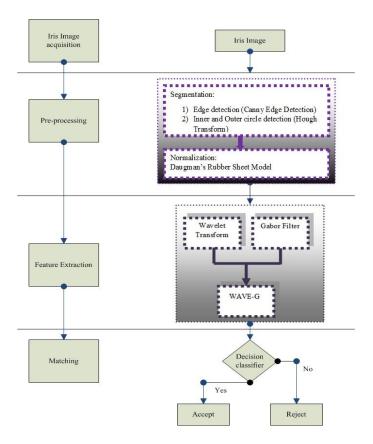


Figure 1. Iris recognition framework

Figure 1 shows the proposed model for iris recognition with significant feature extraction. This model divided into three main levels. First level is having pre-processing steps which are necessary for the desired tasks. Level one is further divided into two sub steps. We get iris image and start performing processing steps which first phase is segmentation where this step is consists of edge detection besides detecting the inward and external restrictions of the iris. In second level, select parameters for feature extraction then apply Gabor filter and wavelet transform and combined these two methods results according to the Haralick features scheme.

#### 2.1. Wavelet: a suitable candidate for texture analysis

A near assessment of iris clone exposes the situation fertility in consistency. The grain material are sophisticated by the pendant. A portion since that, it involves of arbitrary distinguishings similar centrifugal stations, concentric channels, tincture points, and catacombs. These characteristics have unalike regularities, which canister is preeminent defined through a multi-determination inquiry (MRA) method [4], [5]. The remote wavelet change is MRA coordination, which dismiss brand the iris duplicate in unalike positioning plus vanquishing. Wavelet transmute is a multi-persistence performance, which consumes inductee effective demonstrations in numeral image allowance. The wild wavelet change is a precise algorithm intentional to chance a waveform or indicator in the spell domain into a cataloguing of measures grounded on an orthogonal source of unimportant determinate waves, or wavelets. These renovate dismiss be simply protracted to multidimensional signals, such as pictures, wherever the time sphere is traded through the interplanetary ground. It is consequential since a finitely fashioned, orthogonal MRA. In the bonds agreed there, one indicates a range scale J with test group rate of  $2^J$  per unit pause, and strategies the specified

indicator f onto the planetary V<sub>J</sub>; in theory by withdrawing the scalar produces. The procedure is reckless and has a low compactness gradation. The assembly adapts the assemblies to yield its iris piece codes [6], [7]. In direct division disengaged wavelet transmute [8]-[13]. The picture are essentially disintegrated i.e., separated into four sub-bands and unsympathetically sub-sampled by smearing detached wavelet convert as exposed in Figure 2. These sub-bands branded LH<sub>1</sub>, HL<sub>1</sub>, and HH<sub>1</sub> distinguish the supreme ruler wavelet measures i.e., aspect pictures whereas the sub-band LL<sub>1</sub> look like to grainy smooth amounts i.e., evaluation image. Towards gain the subsequent patchy near of wavelet quantities, the sub-band LL<sub>1</sub> unaided is added fragmented and frowningly investigated. These products in two-level wavelet disintegration as shown in Figure 2(b). Correspondingly, to accomplish auxiliary putrefaction, LL<sub>2</sub> will stand secondhand [14]. This extension remainders tray selected concluding sovereign is anxious. The integrities or deformed amounts in scheming and sub-band images are the essential assemblies, which stand revealed now as apposite aimed at surface examination and awareness. As palmtop-touches or function-surfaces need non-unbroken steely level dissimilarities, they are statistically patented by the structures in appraisal and aspect pictures. In additional words, the prices in the represent-posse images or their consolidations or the consequent structures from these ensembles distinctively explain a consistency. The structures grown after these wavelet rehabilitated images are shown to be used for consistency tagging [15]–[17].

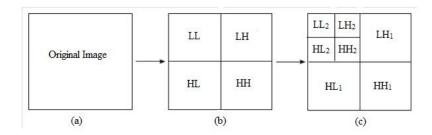


Figure 2. Two level decomposition by using discrete wavelet transform for (a) new image, (b) first section breakdown, and (c) second section breakdown [18]

#### 2.2. Fast wavelet transform

The fast wavelet transform (FWT) is mathematical which is technique planned for the signal or wave. In the time range, it will convert obsessed by a sequence of measurements constructed scheduled an orthogonal basis of small finite waves or wavelets. The convert can be effortlessly protracted to multi-dimensional signals such as pictures. This algorithm was presented in [19], hypothetical base of this algorithm is to produce the orthogonal MRA finitely. The Mallat algorithm is an average method for distinct wavelet convert which is branded as a two-channel sub-group coder. Therefore, this conversion comprises of two parts as [20]: i) The disintegration procedure jumps with signal s, next determines the harmonizes of  $A_1$  and  $D_1$  and before persons of  $A_2$  and  $A_3$  and  $A_4$  and  $A_4$  and  $A_4$  and  $A_4$  and  $A_4$  and then complete the synchronizes of  $A_4$  and  $A_4$  analyses individuals of  $A_4$  and so on [21]. In the multi-determination agenda, an orthogonal wavelet twitches through the ascending task  $\alpha$  then the wavelet purpose  $\alpha$ . One of the essential relatives stands the identical-ruler relative which is defined in (1) and the wavelet function is in (2) as [22], [23].

$$\phi(x) = \sum_{k=-N}^{N} (w_n \phi(2x - k)) \tag{1}$$

$$\psi(x) = \sum_{k=-N}^{N} ((-1)^k w_{1-k} \psi(2x - k))$$
(2)

Altogether the filters used in dissimilar wavelet convert and inverse distinct wavelet changes are confidentially associated to the system given in (3) [24].

$$(w_n)_{n\in\mathbb{Z}} \tag{3}$$

Obviously, if  $\varphi$  is trimly reinforced the arrangement  $w_n$  is determinate then canister be regarded as a filter. The strainer w which is called the clambering filter is having some belongings such as fixed compulsion response (FCR), distance 2N, amount 1, standard  $\frac{1}{\sqrt{2}}$  and a low-slung permit filter. From filter w outline four FIR filters of length 2N and of norm 1 equipped as unprotected in Figure 3 [25]–[27].

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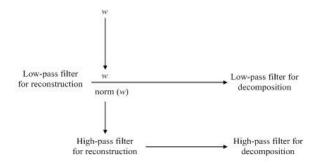


Figure 3. How to compute four filters

Assumed a signal s of distance N the detached wavelet adapts comprises of  $log_2N$  periods at highest. Inaugural since s, the first step harvests two sets of dimensions: approximation amounts  $cA_1$  and feature numbers  $cD_1$ . These trajectories are prolonged by convolvings through the low-pass strainer for calculation and with the high-pass strainer for feature are monitored by dyadic devastation. Additional accurately the first step is shown in Figure 4 [28], [29].

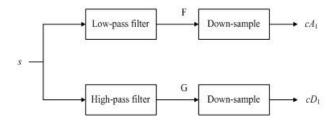


Figure 4. First phase of discrete wavelet transform

The extent of every strainer is identical to 2L. The product of convolving a dimension N indicator with a dimension 2L filter is N+2L-1. Hence, the indicators F and G are of measurement N+2L-1. Later down-selection by 2, the constant vectors  $cA_1$  and  $cD_1$  are of length  $\left\lfloor \frac{N-1}{2} + L \right\rfloor$ . Then succeeding step splits the calculation numbers  $cA_1$  in two portions consuming the same arrangement transaction s by  $cA_1$  and manufacturing  $cA_2$  and  $cD_2$  which displays in Figure 5 [30], [31].

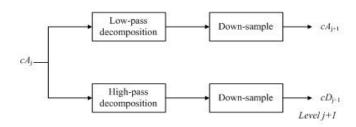


Figure 5. 1-D separate wavelet transform for disintegration stage

Equally, preliminary from  $cA_j$  and  $cD_j$  the opposite disconnected wavelet convert renovates  $cA_{j-1}$ , upsetting the decomposition phase through implanting zeros and convolving the consequences by the rebuilding filters [32]. Figure 6 shows the renovation step of 1-D separate wavelet transform.

Figure 6. 1-D discrete wavelet transform for renovation phase

On behalf of images, a comparable algorithm is conceivable for 2-D wavelets and ascending meanings increased after 1-D wavelets by tensorial construction. This generous of 2-D discrete wavelet converts indications to a putrefaction of estimate quantities at equal j in four workings: the evaluation at level j+1 and the details in three alignments such as straight, upright and diagonal [33]. The discrete wavelet transmute of a discrete signal  $X = [x[0], ..., x[N-1]]^T$  is the process of getting the coefficients as in (4) and (5) [34].

$$W_{\varphi}[j_0, k] = \frac{1}{\sqrt{N}} \sum_{m=0}^{N-1} x[m] \varphi_{j_0, k}[m] = \frac{1}{\sqrt{N}} \sum_{m=0}^{N-1} x[m] 2^{j_0/2} \varphi[2^{j_0} m - k], \forall_k$$
 (4)

$$W_{\Psi}[j,k] = \frac{1}{\sqrt{N}} \sum_{m=0}^{N-1} x[m] \Psi_{j,k}[m] = \frac{1}{\sqrt{N}} \sum_{m=0}^{N-1} x[m] 2^{j/2} \Psi[2^{j}m - k], \forall_{k} \text{ and all } j > j_{0}$$
 (5)

While the source scaling and wavelet meanings are correspondingly formulated in (6) as monitors [35].

$$\begin{cases} \varphi_{j,k}[m] = 2^{j/2} \varphi[2^{j}m - k] \\ \Psi_{j,k}[m] = 2^{j/2} \Psi[2^{j}m - k] \end{cases}$$
 (6)

Remember that both the scrabbling and wavelet meanings can be prolonged in terms of basis mounting purposes of the next higher purpose which shows in (7) as [35], [36].

$$\begin{cases}
\varphi[m] = \sum_{l} h_{\varphi}[l] \sqrt{2} \varphi[2m - l] \\
\Psi[m] = \sum_{l} h_{\Psi}[l] \sqrt{2} \varphi[2m - l]
\end{cases}$$
(7)

Currently seeing the algorithm as a wild origination to get the constants  $W_{\varphi}[j,k]$  and  $W_{\psi}[j,k]$  of different scales j. Consider first the climbing function  $\varphi[m]$ . Trading m by  $2^{j}m - k$  (scaled by  $2^{j}$  and decrypted by k), then the mounting function develops like in (8) [37], [38].

$$\varphi[2^{j}m - k] = \sum_{l} h_{\varphi}[l] \sqrt{2} \varphi[2(2^{j}m - k) - l] = \sum_{l} h_{\varphi}[l] \sqrt{2} \varphi[2^{j+1}m - 2k - l]$$
 (8)

Now the replacement of n=2k+l and l=n-2k, then (8) becomes just as (9) [38].

$$\varphi[2^{j}m - k] = \sum_{n} h_{\varphi}[n - 2k]\sqrt{2}\varphi[2^{j+1}m - n]$$
(9)

Likewise, the wavelet purpose can also be long-drawn-out just as in (10) [39].

$$\Psi[2^{j}m - k] = \sum_{n} h_{\Psi}[n - 2k]\sqrt{2}\varphi[2^{j+1}m - n]$$
(10)

This wavelet purpose which labeled in (10) is indistinguishable to the unique secondhand in comparison 5. So, supernumerary (10) into (5) then it converts like (11) [40].

$$W_{\Psi}[j,k]: \frac{1}{\sqrt{N}} \sum_{m=0}^{N-1} x[m] 2^{\frac{j}{2}} \Psi[2^{j}m - k] = \frac{1}{\sqrt{N}} \sum_{m=0}^{N-1} x[m] 2^{\frac{j}{2}} \left[ \sum_{n} h_{\Psi}[n - 2k] \sqrt{2} \varphi[2^{j+1}m - n] \right]$$
$$: \sum_{n} h_{\Psi}[n - 2k] \left[ \frac{1}{\sqrt{N}} \sum_{m=0}^{N-1} x[m] 2^{(j+1)/2} \varphi(2^{j+1}m - n) \right]$$
(11)

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The appearance  $\frac{1}{\sqrt{N}}\sum_{m=0}^{N-1}x[m]2^{(j+1)/2}\varphi(2^{j+1}m-n)$  materializes to be the wavelet change for the quantity of scale j+1 in (12) [41].

$$W_{\psi}[j+1,k] = \frac{1}{\sqrt{N}} \sum_{m=0}^{N-1} x[m] 2^{(j+1)/2} \varphi(2^{j+1}m - n)$$
(12)

Consequently, a recursive relation among the wavelet transforms constants of two uninterrupted scale heights j and j+1 shown in (13) and the similar is true to the mounting meaning which is also exposed in (14) [39]–[41].

$$W_{\Psi}[j,k] = \sum_{n} h_{\Psi}[n-2k]W_{\Psi}[j+1,n] \tag{13}$$

$$W_{\varphi}[j,k] = \sum_{n} h_{\varphi}[n-2k]W_{\varphi}[j+1,n]$$
(14)

Once associating (13) and (14) with a discrete difficulty then we get the consequence as a form of (15) as [42]-[44].

$$y[k] = h[k] * x[k] = \sum_{n} h[k - n]x[n]$$
(15)

Accordingly, the wavelet renovate constants  $W_{\varphi}[j,k]$  and  $W_{\Psi}[j,k]$  at the  $j^{th}$  scale canister stand attained from the quantities  $W_{\varphi}[j+1,k]$  and  $W_{\Psi}[j+1,k]$  at the  $(j+1)^{th}$  scale by two belongings such as complication with time reversed  $h_{\varphi}$  or  $h_{\Psi}$  and sub-sampling to get every other illustrations in the convolution. Therefore, wavelet transmute and scaling meaning will become the shape as given in (16) as [45], [46].

$$\begin{cases}
W_{\Psi}[j,k] = h_{\Psi}[-n] * W_{\varphi}[j+1,n]|_{n=2k,k\leq 0} \\
W_{\varphi}[j,k] = h_{\varphi}[-n] * W_{\varphi}[j+1,n]|_{n=2k,k\leq 0}
\end{cases}$$
(16)

Created on (16), all wavelet then scrabbling numbers  $W_{\Psi}[j,k]$  and  $W_{\varphi}[j,k]$  of a agreed indication X can be gotten recursively from the constants  $W_{\Psi}[j,k]$  and  $W_{\varphi}[j,k]$  at the highest determination level j=J with all concentrated information and the N data points x[m] (m=0,...,N-1) straight tried after the signal x(t). As an associate of space  $V_J$ , these separate samples can be written as a rectilinear mixture of the climbing source purposes  $\varphi_{I,k}[m]$  which shows in (17) as [47], [48]. Figure 7 present wavelet transform.

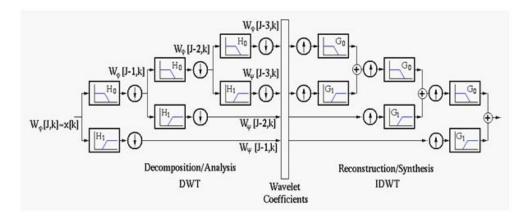


Figure 7. Diagram of fast wavelet transform with its operational [49]

# 2.3. Gabor filter

In image processing a Gabor filter is a linear filter used for texture analysis. As a linear filter, it reflects the account as monitors:

*Process*[(time varying input signals) → (output signals)] Subject to linearity

Where linearity is a stuff which revenues that it can be explicitly characterized as a traditional line. Conferring to the linear function f(x) that function must indulges the two effects. First property is the conserving stuff as  $\{f(x+y)=f(x)+f(y)\}$  and the second is equality of degree 1 which is  $\{f(ax)=af(x), \forall a\}$  [38]. It mostly scrutinizes whether nearby any explicit regularity pleased trendy the pictures in plain directives in a constrained division everywhere the theme or pitch of investigation. In the latitudinal area, a 2D Gabor filter is a Gaussian Kernel function altered by a sinusoidal plane wave. Its impulse response is well-defined by a sinusoidal wave which augmented by a Gaussian function [50], [51]. Owing to the development intricacy stuff, the transformation of a Gabor filter's obligation rejoinder is the convolution of the choral determination which is identified as sinusoidal meaning and the Gaussian function. The filter has a material and a make-believe essential on behalf of orthogonal remits [52], [53]. These two workings may be designed into a multifarious number or used independently. As a complex method of Gabor filter is exposed in Equivalence 18 as:

$$g(x, y; \lambda, \theta, \Psi, \sigma, \gamma) = exp\left(-\frac{\dot{x}^2 + \gamma^2 \dot{y}^2}{2\sigma^2}\right) exp\left(i\left(2\pi \frac{\dot{x}}{\lambda} + \Psi\right)\right)$$
(17)

Material and invented parts of (18) revealed in (19) as:

$$\begin{cases} g(x,y; \lambda, \theta, \Psi, \sigma, \gamma) = exp\left(-\frac{\dot{x}^2 + \gamma^2 \dot{y}^2}{2\sigma^2}\right) \cos\left(2\pi \frac{\dot{x}}{\lambda} + \Psi\right), Real \ part \\ g(x,y; \lambda, \theta, \Psi, \sigma, \gamma) = exp\left(-\frac{\dot{x}^2 + \gamma^2 \dot{y}^2}{2\sigma^2}\right) \sin\left(2\pi \frac{\dot{x}}{\lambda} + \Psi\right) \ Imaginary \ part \end{cases}$$
(18)

Where:

- $x' = x\cos\theta + y\sin\theta$  and  $y' = y\cos\theta x\sin\theta$  are the components
- λ Characterizes the wavelength of the sinusoidal inspiration
- $\theta$  Denotes the alignment of the usual to the equivalent stripes of a Gabor meaning
- $\psi$  is the stage offset,  $\sigma$  is the standard deviation of the Gaussian covering
- γ is the longitudinal facet relation and requires the ellipticity of the sustenance of the Gabor function

A fixed of Gabor filters with diverse occurrences and bearings can stay cooperative aimed at excavating convenient structures after an image (Haghighat *et al.*, 2013). In the distinct field, 2D Gabor filters are given by [54], [55].

$$\begin{cases}
G_c[i,j] = Be^{-\frac{(i^2+j^2)}{2\sigma^2}}\cos(2\pi f(i\cos\theta + j\sin\theta)) \\
G_s[i,j] = Ce^{-\frac{(i^2+j^2)}{2\sigma^2}}\sin(2\pi f(i\cos\theta + j\sin\theta))
\end{cases}$$
(19)

Where B and C are regulating factors to be determined, f defines the frequency. By changing  $\theta$  is used for consistency orientation in a certain bearing while through the variation of n, dissimilarity [56]–[58].

#### 3. EXPERIMENTAL RESULTS AND COMPARISON

Assessing these presentations of biometric technique is a problematic production. For the determination of judgment; we contrivance these approaches interpretation to the reproduced documents. To assistant their recital, we cast-off three categories of statistics which are UBIRIS, CASIA, and MMU. Iris database covers 280 eye images from 28 objects and each person has 10 images of eye. Altogether hearings were achieved by MATLAB version R2010b on the core processor. We use the usual technique to generate and normalize iris extents and use the amalgamation of three methods acknowledged overhead to mine the quantity. Consequently, we individual inspect and companion the exactness and computational application of chin extraction. Subsequently piece construct, we usage hybrid classifier for corresponding period (i.e) hard-to-test faults (HTTF) and false acceptance rate (FAR)/false rejection rate (FRR) are used for appraising the outcome.

## 4. ACCURACY RATE OF PREPROCESS SECTION

The precision of the planned technique is charted in Table 1 shows that the algorithm mechanism well even with the retinal images with complaints. From Table 1, Table 2 and Figure 8 consequence the result of preprocess constructed on the result the offered hybrid algorithm brand enhancements in the

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preprocess segment. The development involve of changed dataset like UBIRIS, CASIA, MMU the development result show that the anticipated technique upgrading in the preprocess sector also.

Table 1. Preprocess of different data set result

Database	Total Images	Extracted successfully	Accuracy (%)
UBIRIS	280	97%	98.5%
OBIRTIO		,,,,	,
CASIA	550	98%	98.9%
MMU	450	99%	99%

Table 2. Preprocess result of all technique

1 4010 2.110 0100000 100411 01 411 100111110 400					
Database	Total images	Extracted successfully	Accuracy (%)		
ANN	280	97%	98.5%		
Block sum algorithm	550	98%	98.9%		
SVM	450	99%	99%		
Proposed technique					

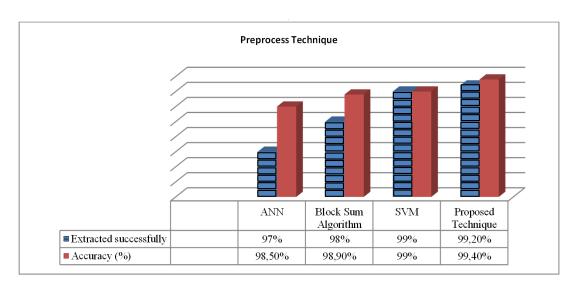


Figure 8. Overall result of preprocess

Table 3. Experimental results

Technique name	FAR/ FRR	Overall accuracy rate
ANN	4/6	97%
Block sum algorithm	3.2/3.1	97.5%
SVM	3/4	98%
Proposed technique	6/5	98.2%

False acceptance rate (FAR): is the possibility of the organization which imperfectly matches the input image with the no matching prototype in the database. It plaid the improper image in the database in situation of assessment gage, gamble the individual is pretender in actual then the identical notch is multifaceted than inception besides he is saline as unassuming that increases the FAR plus hereafter the recital similarly be contingent upon the variety of inception rate.

False dismissal rate (FRR): The position that the scheme plunges to extricate a contest amongst the effort outline and a different decoration in the record. It arrangements the out of a hundred of effective involvements which are wrongly rejected. Table 4 and Figure 9 show the result rendering to the result the suggested technique 98.20% overall result of the IRIS recognition as associate to the other algorithms.

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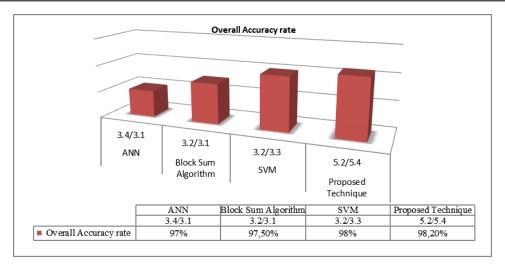


Figure 9. Over all result

#### 5. CONCLUSION

In this article, we have debated feature abstraction and classifier of iris recognition by hybrid algorithm. The system be contingent largely on two stages, iris preprocess and iris classifier method. We need logical those transform preceding the iris pictures for encounter obtainable the appreciation proportion and exactitude. Outcomes of this test have publicized that the truthfulness in acknowledgment by fusion procedure is improved than artificial neural network (ANN), block sum algorithm, and support-vector machine (SVM). Likewise Crossbreed classifier i.e. mixture of ANN and FAR/FRR stay cast-off for matching whicheternally image is predictable or rejected. FAR and FRR in section by individual numerous approaches equally revealed in beyond diagram. Therefore, the planned process offers improved precision and appreciation amount those relative systems.

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