

Skin Texture Analysis Based On New GLCM & Modified Thresholding

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Abstract: Skin surface examination is one of the testing issues in the field of medicinal analysis. Different sorts of skin illnesses are influencing human life like skin dryness, parasite, & hypersensitive side effects. The target of this paper is to examine skin sickness utilizing surface investigation of skin picture & by contrasting test picture with a characterized picture or reference picture. Support Vector Machine (SVM) based GLCM has been connected for perception of highlights. Preparing of SKIN TEXTURE pictures is one of the parts of this field. MATLAB is being utilized for the outline of the proposed framework, to make it easy to use a GUI is being created for front end use & out of sight calculation works, outcomes found are great as far as precision & time delay.

Keywords: STA-Skin Texture Analysis, GLCM-Gray level co-occurrence matrix, SVM-Support Vector Machine, FAR-False acceptance Rate, FRR-False Rejection rate

I-INTRODUCTION

Surface examination is one of the components in picture processing used to investigate pictures that are caught by imaging gadgets on human skin. Typically human skin surface has distinctive sorts like smooth, dryness which is happened in light of individual human nourishment propensities, living condition, hereditary & so on. Skin surface shifts relying upon its age also. Significant properties of skin are harsh, smooth, irregular & standard. Accompanying are some of the regular skin sicknesses,

- a) Cancer Skin Disorders – it is a result of disease infection
- b) Viral Skin Disease - Disorders caused by infections, for example, shingles, mouth blisters & measles
- c) Bacterial Skin Diseases – It is caused by bacterial contaminations, for example, skin break out & folliculitis

- d) Fungal Skin Diseases – Disorders, for example, ringworm, athlete's foot & yeast contaminations





	
Cancer	Fungal
	
Bacterial	Viral

Table 1 possible skin types in case of disease

II-METHODOLOGY

The proposed work has been animated & arranged because of its key hugeness in current conditions. There are a couple of ways & frameworks to expel information from a therapeutic picture to help masters. Thresholding is an amazingly prior bit of it, there are various strategies for Thresholding [7-12]. Affirmation of human skin diseases from SKIN TEXTURE pictures is a critical errand, it is extremely possible that it can be disregarded here & there; in this way proposed computation is thought of as a correct game plan. This technique uses different Otsu Thresholding close by GLCM. MATLAB is being used for the layout of the proposed structure, to make arrangement straightforward a GUI is being delivered for front end use & beyond anyone's ability to see proposed count works, results found are incredible in regards to exactness & time delay.

Figure 1 showed up underneath the proposed work process diagram here the whole work is being disconnected in three imperative tasks

- Preprocessing
- Database Preparation
- Recognition

preprocessing is done by using Otsu Thresholding where photo had distributed taken after by breaking down & enlarging. whole methodology is shown here in flowchart given underneath

3.1.1 PREPROCESSING: Pre-dealing with is a basic anticipated that endeavor would be done in skin diseases affirmation system plot. Pre-planning contain for two phases

- Median channel & Equalization
- Diseased skin zone withdrawal
- Thresholding

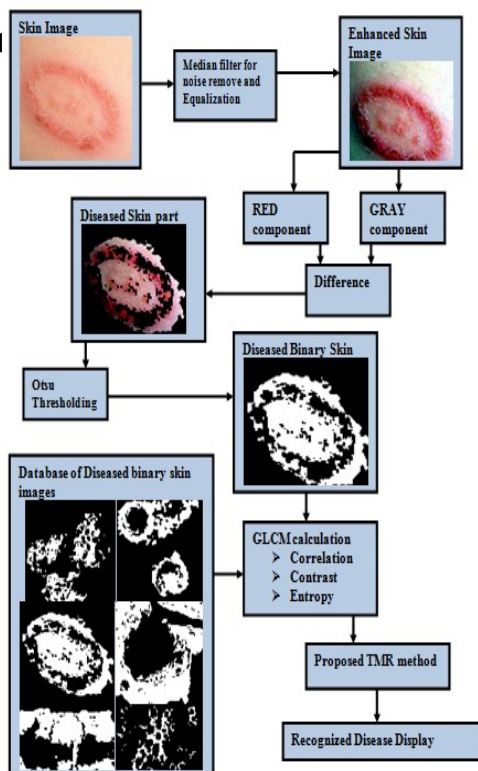


Figure 1 Proposed work block diagram

middle channel is a nonlinear computerized sifting method, regularly used to expel commotion from a picture or flag. Such commotion lessening is an average pre-preparing advance to enhance aftereffects of later handling (for instance, edge location on a picture). Middle sifting is broadly utilized as a part of advanced picture preparing in light of fact that, under specific conditions, it jelly edges while evacuating commotion.

To illustrate, utilizing a window size of three with one section quickly going before & following every passage, a middle channel will be connected to accompanying basic 1D flag:

$$x = (2, 80, 6, 3).$$

Along se lines, middle separated yield flag y will be:

$$y_1 = \text{med}(2, 2, 80) = 2,$$

$$y_2 = \text{med}(2, 80, 6) = \text{med}(2, 6, 80) = 6,$$

$$y_3 = \text{med}(80, 6, 3) = \text{med}(3, 6, 80) = 6,$$

$$y_4 = \text{med}(6, 3, 3) = \text{med}(3, 3, 6) = 3,$$

i.e. $y = (2, 6, 6, 3).$

Histogram adjust is a technique for changing picture forces to enhance separate. distinction can be extended using histogram broadening. In this instructional exercise we will see that how histogram leveling can be used to redesign separate. histogram evening out, it must know two basic thoughts used as a piece of adjusting histograms. se two thoughts are known as PMF & CDF. PMF (probability mass limit) of extensive number of pixels in this photo & CDF (add up to distributive limit)

To test running with code, hist eq.m, sort g = hist_eq('xyz.bmp'); Histogram evening out is moreover consolidated with MATLAB.

Wiped out skin recognizing verification Skin area is route toward finding skin-shaded pixels & zones in a photo or a video. This method is usually used as a preprocessing development to find districts that possibly have human faces & members in pictures, three guideline parameters for seeing a skin pixel are RGB (Red, Green, Blue), HSV (Hue, Saturation, Value) & YCbCr (Luminance, Chrominance) shading models. proposed estimation changes over entire picture in a two dimensional system in which portion & line measure is described by width & stature of photo independently. Once photo is isolated, each section contains a pixel of photo. ARGB shade of that particular pixel is settled. ARGB regard recuperated from photo for each pixel is a 32-bit regard. Thusly to expel each sub-regard i.e. red, green, blue & alpha we right move this motivator by 24 bit with a particular ultimate objective to get estimation of alpha.

Proposed Thresholding: An extraordinary Thresholding is relied upon to pick a tasteful edge for diminish level in order to expel blood stream (white an area) .i.e. re is no part for white range should be in no blood zone & no blood zone similarly shouldn't have any part for blood zone. With everything taken into account, decision for a fitting Thresholding figuring depends, all things considered, on sort for pictures & application domains. Otsu Thresholding count [3] was attempted & found to give incredible Thresholding realizes demand to SKIN TEXTURE pictures & was, thusly, picked.

Otsu figuring is nonparametric & unsupervised system for modified edge decision. Changing over a greyscale picture to monochrome is a run of mill picture planning errand. Otsu's methodology, named after its pioneer Nobuyuki Otsu, is one of various binarization computations, Otsu's Thresholding procedure incorporates rehashing through all possible edge regards & figuring a measure of spread for pixel levels each side of point of confinement, i.e. pixels that ei r fall in bleeding edge or establishment. fact is to find point of confinement regard where entire of nearer view & establishment spreads is at its base. figuring will be shown using clear 6x6 picture exhibited as takes after.

DERM database: <http://dermnetnz.org/is> WMO(world helpful Organization) standard site which gives an enormous number of skin pictures with different afflictions proposed work picked 20 skin pictures for harm, 20 skin pictures for bacterial, 20 skin pictures for infectious & 20 skin pictures for viral total 80 pictures in DERM database



Figure 2 Bacterial image in DERM database



Figure 3 Cancerous image in DERM database



Figure 4 Fungal image in DERM database



Figure 5: Viral image in DERM database

RECOGNITION: SKIN TEXTURE organizing is a one-to-many planning framework that takes a gander at a test SKIN TEXTURE picture against all format SKIN TEXTURE pictures in DERM database to choose character remembering true objective to test SKIN TEXTURE. Recognizing evidence to test picture is done by discovering picture in database that has most amazing resemblance with test picture. Recognizing verification procedure is a 'closed' test, which suggests sensor takes a recognition remembering ultimate objective to a man that is known to be in database. Guinea pig's (institutionalized) GLCM features are appeared differently in relation to o r GLCM incorporates into structure's database & a comparability score is discovered remembering true objective to each examination. se similarity scores are n numerically situated in a falling solicitation. Rate remembering true objective to times that most amazing likeness

score is correct match to all individuals is insinuated as 'top match score.' If any of best r equivalence scores identifies with guinea pig, it is considered as a correct match similar to consolidated match. Level of times one of those likeness scores is correct match surprisingly is implied as 'Add up to Match Score', it is a curve is rank n versus level of right recognizing confirmation, where rank n is number of best closeness scores reported.

Diminish Level Co-Occurrence Matrix (GLCM): A quantifiable methodology for examining surface that considers spatial relationship of pixels is diminish level co-occurrence matrix (GLCM), generally called diminish level spatial dependence grid. GLCM limits depict surface of a photo by registering how every now & again joins of pixel with specific regards & in a predefined spatial relationship occur in a photo, making a GLCM, & reafter expelling accurate measures from this system. (surface channel limits, depicted in Texture Analysis can't give information about shape, that is, spatial associations of pixels in an image.)After you make GLCMs, using graycomatrix, you can get a couple of estimations from m using graycoprops. se estimations give information about surface of a photo. going with table records bits of knowledge.

Statistic	Description
Contrast	Measures neighborhood varieties in dim level co-event network.
Correlation	Measures joint likelihood event of predetermined pixel sets.
Entropy	Gives total of squared components in GLCM. O rwise called consistency or rakish second minute.

Table 2 Proposed GLCM parameters

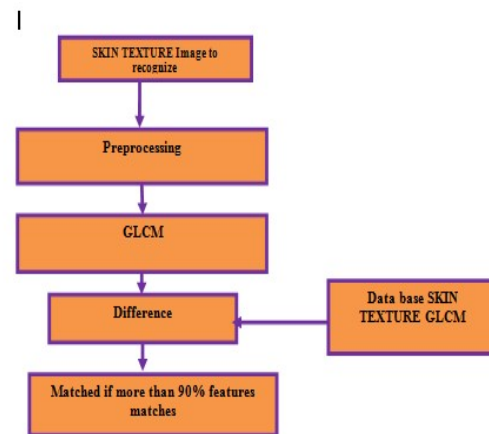


Figure 6: Proposed Recognition method
Proposed method has two major parts

- Pre-processing
- Identification

PRE-PROCESSING: Proposed work is basically image quality enhancement through median filter & disease binary image generation using Otsu Thresholding.

Let $X_{i,j}$ is test skin image with $M \times N$ size,
 $i=1, 2, 3, \dots, M$
 $j=1, 2, 3, \dots, N$
 'j' & 'i' are pixel position

$$Z_{i,j} = \sum_{i=1}^{n-1} \sum_{j=1}^{m-1} \{x(n-i, m-j)\} * (C_{i,j})$$

where $(C_{i,j})$ are filter coefficients

$Z_{i,j}$ is enhanced filtered image

$$Z_{red} = Z_{i,j}(:, :, 1)$$

$$Z_{green} = Z_{i,j}(:, :, 2)$$

$$Z_{blue} = Z_{i,j}(:, :, 3)$$

$$Z_{gray} = \left(\frac{Z_{red} + Z_{green} + Z_{blue}}{3} \right)$$

$$Y_{i,j} = |Z_{gray} - Z_{red}|$$

$Y_{i,j}$ is disease area skin image of test human

cnt1, cnt2, cnt3, cnt4 & cnt5 are count values that are been assigned to get count value so we could know quantity of pixel values that lies in specific range



```

if(Yij > 0 Yij < 51)
cnt5 = cnt5 + 1;
elseif(Yij > 50 Yij < 101)
cnt4 = cnt4 + 1;
elseif(Yij > 100 Yij < 151)
cnt3 = cnt3 + 1;
elseif(Yij > 150 Yij < 201)
cnt2 = cnt2 + 1;
elseif(Yij > 200 Yij < 255)
cnt1 = cnt1 + 1;

```

$$Pcnt1 = \frac{cnt1 * 100}{cnt1 + cnt2 + cnt3 + cnt4 + cnt5}$$

$$Pcnt2 = \frac{cnt2 * 100}{cnt1 + cnt2 + cnt3 + cnt4 + cnt5}$$

$$Pcnt3 = \frac{cnt3 * 100}{cnt1 + cnt2 + cnt3 + cnt4 + cnt5}$$

$$Pcnt4 = \frac{cnt4 * 100}{cnt1 + cnt2 + cnt3 + cnt4 + cnt5}$$

$$Pcnt5 = \frac{cnt5 * 100}{cnt1 + cnt2 + cnt3 + cnt4 + cnt5}$$

```

if(pcnt1 > 10)
max = 255
elseif((pcnt1 + pcnt2) > 10)
max = 200
elseif((pcnt1 + pcnt2 + pcnt3) > 10)
max = 150
elseif((pcnt1 + pcnt2 + pcnt3 + pcnt4) > 10)
max = 100
else
max = 50

```

THRESHOLDING: T is Thresholding level which decided according to Otsu Thresholding method
 $T = 50 * \sqrt{2 * \log(Max)}$

$$T_{ij} = \begin{cases} 255 & |Y_{ij}| \geq T \\ 255 * \text{sgn}(Y_{ij}) * \frac{|Y_{ij}|^\gamma}{T^{\gamma-1}} & |Y_{ij}| < T \end{cases}$$

Where $\gamma = \sigma \sqrt{2 * \ln(N)}$
And $\sigma = \text{mean}(Y_{ij})$

Database Preparation: Total DERM_{100x100}(m) where m is 1,2,3...80 SKIN TEXTURE image of normal human of 100x100 pixels.

$$P_{10000 \times 1}(m) = [DERM_{(100 \times 100, 1)}(1), DERM_{(100 \times 100, 1)}(2), \dots \dots \dots DERM_{(100 \times 100, 1)}(n)]$$

Where m= 1, 2, 3.....80, m is number of SKIN TEXTURE's in database

Correlation recognition:

$$r(n) = \sum_{k=0}^n T_{ij}(k)T_{ij}(k-n)$$

'n' is any sample position out of total 10000 samples of T_{ij}

$$r_m(n) = \sum_{k=0}^n T(k)P(k-n, m)$$

$$S = \sum_{n=0}^{10000} r(n)$$

$$S_m = \sum_{n=0}^{10000} r_m(n)$$

$$f_m = |S_m - S|$$

$$(Val1, K1) = \text{Min}(f_m)$$

Hard Thresholding

$$mch = \begin{cases} 1 & \text{if val1} < 10\% \text{ of max} \\ 0 & \text{o rwise} \end{cases}$$

Contrast recognition:

$$s = \sum_{i,j=0}^{n,m} (P_{i,j} - T_{i,j})^2$$

$$(val1, K2) = \text{min} \left\{ \frac{s}{n * m} \right\}$$

Hard Thresholding

$$mch = \begin{cases} 1 & \text{if val2} < 10\% \text{ of max} \\ 0 & \text{o rwise} \end{cases}$$

Entropy Recognition:

$$Ent = \sum_{i=0}^n \sum_{j=0}^m (P_{i,j} - T_{i,j}) \{ \log (P_{i,j} - T_{i,j}) \}$$

$$(val3, K3) = \text{min}\{ent\}$$

Hard Thresholding

$$mch = \begin{cases} 1 & \text{if val3} < 10\% \text{ of max} \\ 0 & \text{o rwise} \end{cases}$$

If mch is '1' n no need for any fur r calculation because SKIN TEXTURE is been matched with any normal human SKIN TEXTURE significantly, But if mch is '0' n not matched.

At last proposed TMR method for final Skin identification

$$K = \begin{cases} K1 & \text{if}(K1 = K2 = K0) \\ K2 & \text{if}(K2 = K0 \text{ or } K2 = K0) \\ K1 & \text{if}(K1 = K0 \text{ or } K1 = K2) \\ K0 & \text{if}(K0 = K2 \text{ or } K0 = K1) \\ K2 & \text{if}(K2 = K0 \text{ or } K2 = K0) \\ K0 & \text{o rwise} \end{cases}$$

At last value of K is most matched image in database which will be consider similar disease as in test image.

III-RESULTS

Figures beneath demonstrates watch comes about after reproduction of proposed take a shot at MATLAB. This figure indicates unique, sectioned, disintegrated, expanded & last identified skin illnesses picture. recreation comes about have been created for six experiments taken from Govt medicinal doctor's facility, Jabalpur.

GUI created in MATLAB: Figure appeared beneath is GUI created with assistance of MATLAB.

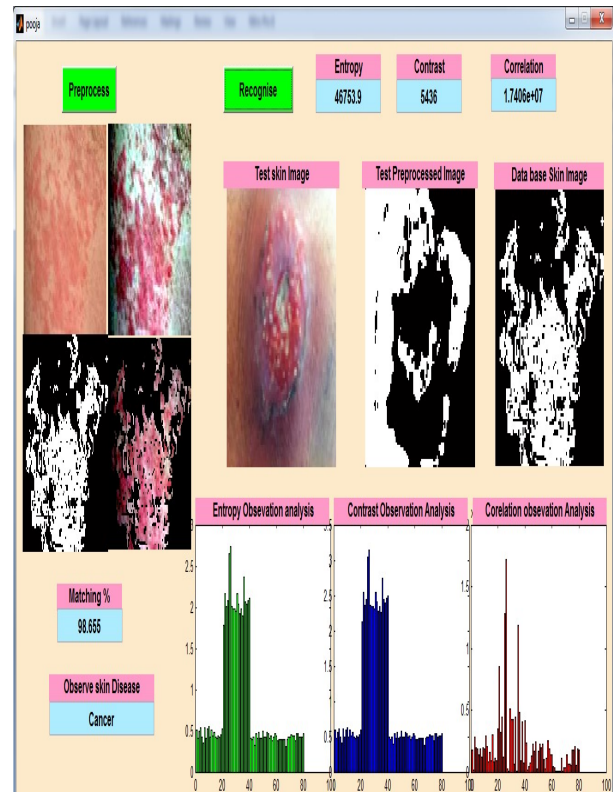


Figure 7 GUI with observe results of GLCM & matching%

In GUI database is taken first & after that it takes test picture in which skin sicknesses must be perceived & when this test picture is stacked, Thresholding & morphological separating is performed on it. This prepared picture is n contrasted & database layouts. In event that test picture is coordinated with any of pictures in database at that point re is no skin sicknesses in test picture & if test picture does not coordinated at that point re is a skin infections in test picture.

Precision: As appeared in figure underneath, it can be plainly watched that proposed work exactness is discovered better as contrast & accessible work.

$$\text{Accuracy} = \frac{\text{Number of right results ei r matched or not matched with test SKIN TEXTURE}}{\text{Total number of attempt}} * 100$$

Coordinating requires edge as in proposed work a hard Thresholding has been utilized, re are two conceivable mistake rates in Thresholding based

coordinating calculations FAR (False Accept Rate) & FRR (False Reject Rate). FAR is figured as a small amount of impostor scores surpassing your edge. FRR is computed as a small amount of certifiable scores falling howl your limit.

$$FAR = \frac{\text{wrong match above threshold}}{\text{all wrong match}}$$

$$FRR = \frac{\text{genuine match below threshold}}{\text{all genuine match}}$$

OBSERVE RESULTS

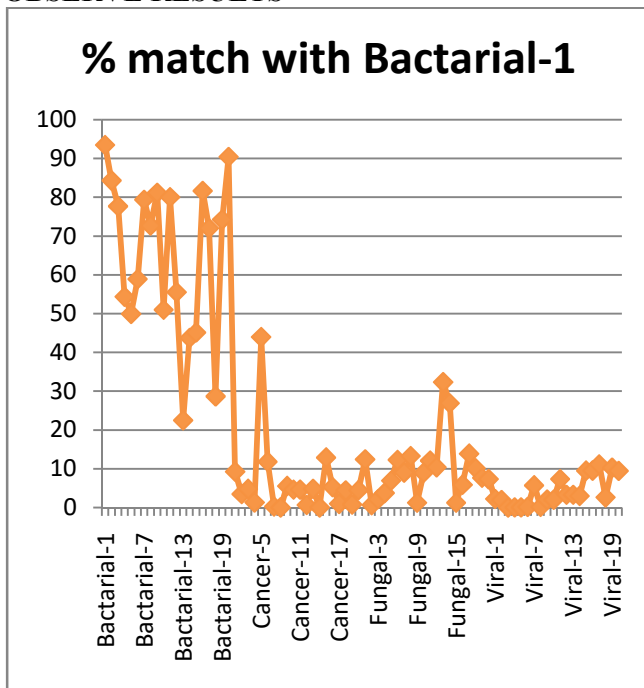


Figure 8 GLCM Results observe with compare with Bacterial disease image

It can be see from table & picture over that when bacterial test skin picture is coordinated with all database skin picture more often than not it gets coordinated with bacterial pictures just & not coordinated with o r skin sick pictures.

Case 1: 2 times out of 20 conceivable right match bacterial picture couldn't coordinated with bacterial picture henceforth.

$$FRR = \frac{2}{20} = 0.1$$

Case 2: 2 times out of 60 possible wrong match bacterial images could right matched with bacterial image hence.

$$FAR = \frac{2}{60} = 0.0333$$

Case 3: 4 times out of 80 possible attempts could get correct matched answer

$$\text{Accuracy} = \frac{(80 - 4)}{80} * 100 = 95\%$$

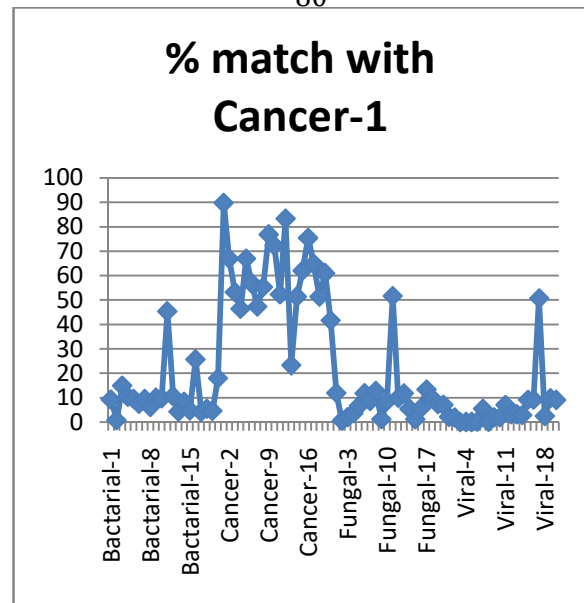


Figure 9 GLCM Results observe with compare with Cancer disease image

It can be see from table & picture over that when malignant test skin picture is coordinated with all database skin picture more often than not it gets coordinated with harmful pictures just & not coordinated with o r skin sick pictures.

Case 1: 1 times out of 20 possible correct match cancerous image could not matched with cancerous image hence.

$$FRR = \frac{1}{20} = 0.05$$

Case 2: 3 times out of 60 possible wrong match cancerous images could right match with cancerous image hence.

$$FAR = \frac{3}{60} = 0.05$$

Case 3: 4 times out of 80 possible attempts could get correct matched answer

$$\text{Accuracy} = \frac{(80 - 4)}{80} * 100 = 95\%$$

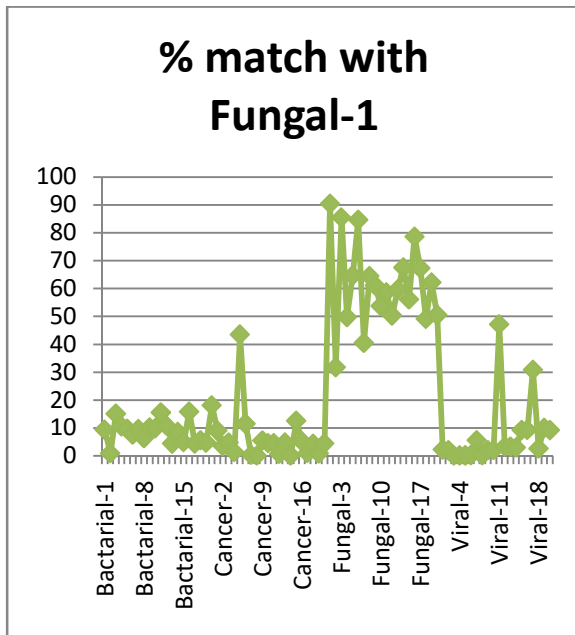


Figure 9 GLCM Results observe with compare with fungal disease image

It can be seen from table & picture over that when parasitic test skin picture is coordinated with all database skin picture more often than not it gets coordinated with contagious pictures just & not coordinated with o r skin unhealthy pictures.

Case 1: 1 times out of 20 conceivable right match parasitic picture couldn't coordinated with contagious picture thus.

$$\text{FRR} = \frac{1}{20} = 0.05$$

Case 2: 2 times out of 60 possible wrong match fungal images could right matched with fungal image hence.

$$\text{FAR} = \frac{2}{60} = 0.0333$$

Case 3: 3 times out of 80 possible attempts could get correct matched answer

$$\text{Accuracy} = \frac{(80 - 3)}{80} * 100 = 96.25\%$$

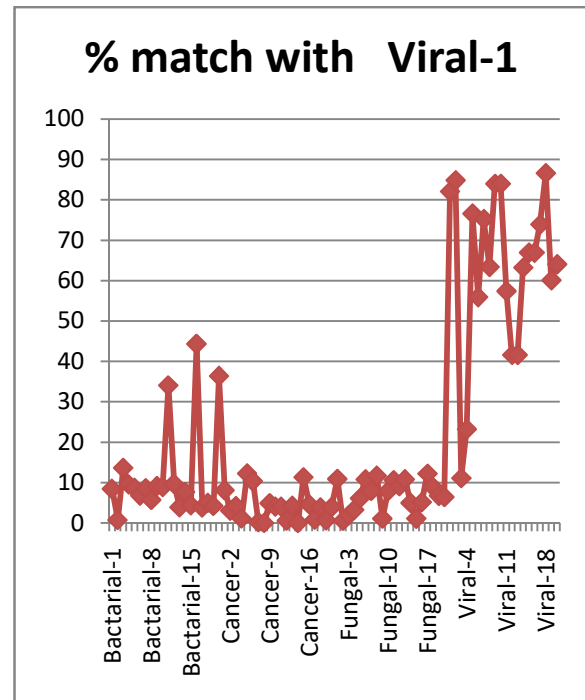


Figure 10 GLCM Results observe with compare with Viral disease image

It can be see from table & picture over that when viral test skin picture is coordinated with all database skin picture more often than not it gets coordinated with viral pictures just & not coordinated with o r skin sick pictures.

Case 1: 0 times out of 20 conceivable right match viral picture couldn't coordinated with viral picture subsequently.

$$\text{FRR} = \frac{0}{20} = 0$$

Case 2: 2 times out of 60 possible wrong match viral images could right matched with viral image hence.

$$\text{FAR} = \frac{2}{60} = 0.01666$$

Case 3: 1 times out of 80 possible attempts could get correct matched answer

$$\text{Accuracy} = \frac{(80 - 1)}{80} * 100 = 98.75\%$$

Form observe results average matching percentage can be calculated as

$$\text{matching} = \frac{95 + 95 + 96.25 + 98.75}{4} = 96.25 \%$$

COMPARATIVE RESULTS

Work By	Accuracy %
Seema Kolkur [1] GLCM-SVM	95
Seema Kolkur [1] GLCM-Neural Network	97.5
Ritesh Maurya [2] GLCM+MSVM	84.2
Proposed GLCM	96.25

Table 3 Comparative Results

FAR, FRR & Accuracy results observe is measure with tic & toc command in MATLAB. From results above it can be clearly observed that proposed work accuracy is found better as compare with available work & it can also be seen that proposed work time to detect skin diseases is less as compare with available work.

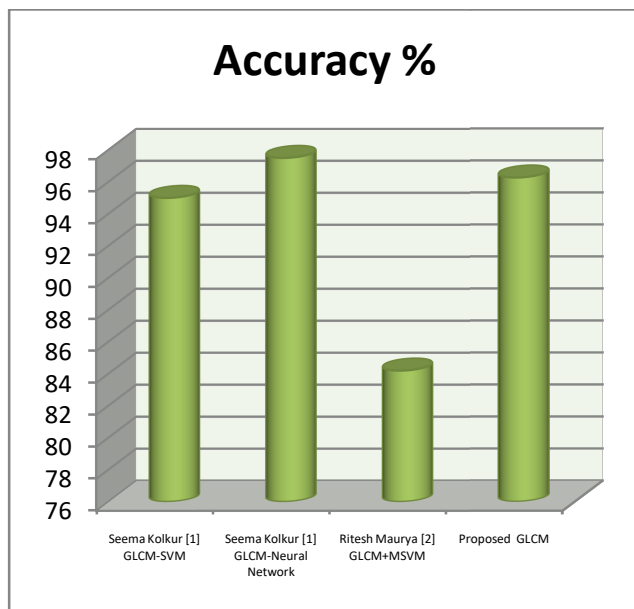


Figure 11 Comparative results of accuracy
IV-CONCLUSION

Proposed method for location of human skin maladies from SKIN TEXTURE pictures & MATLAB is been utilized for outline proposed framework, to make plan easy to use a GUI is been created, outcomes found are great as far as exactness & time delay. DERM (gave by International Medical Union) database is been

utilized for coordinating test SKIN TEXTURE picture to guarantee that it isn't coordinated with any typical human SKIN TEXTURE picture., significant utilization of proposed configuration is to perceive skin ailments out of different SKIN TEXTURE pictures taken for any patient, for precise identification & correct measurement of skin ailments is exceedingly requires in field of rapeutic. Thus main use of proposed configuration is to distinguish Skin sicknesses in SKIN TEXTURE pictures of human. In not so distant future proposed work can be utilized for acknowledgment of o r body parts SKIN TEXTURE & it & be utilized for location of o r illness, as proposed work is effectively perceiving course of blood

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