

HYBRID CLASSIFIER TO CLASSIFY THE FINGER NAIL ABNORMALITIES

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Abstract— Nail diagnosis is a method to predict the possibilities of organ failures and various systemic diseases. Nail abnormalities are considered as the signs of certain diseases in traditional medicines such as Siddha Medicine, Ayurveda, Yunani and Chinese medicine etc. In this paper, the performance of existing techniques such as SVM classifier and KNN classifiers are compared with the proposed method. The metrics precision, recall, F-measure and accuracy are calculated and compared. The 100 images had taken for study and the proposed novel segmentation method gives the best accuracy. The experiment uses 480 (increase the dataset) images of eight types of abnormalities. 70% of images were used for training and 30% of images were used for testing. (Discuss the performance measure)

Keywords— Finger nail image analysis; KNNwithSVM classifier; KNN classifier; SVM classifier; finger nail disease prediction;

I. INTRODUCTION

In medical Imaging diagnosis methods are used to measure the severe of the disease and for further treatment. Some may invasive and others may non-invasive techniques. Mostly, the diseases can be identified by symptoms and external appearances. Generally, the medical practitioners check the eye, tongue, nail and palm abnormalities as the indication of the diseases and they prefer further diagnosis method or investigations. Here, the nail diagnosis method is considered for the research work using medical image processing method, which contributes to tele-medicines, mass-screening and tele-dermatology fields. Nail abnormalities are the signs of the systemic diseases happening in the human body. The abnormalities of finger nails are the signs for various disease and organ failures or for severe systemic diseases. Nail abnormalities are classified by it color, texture, size and shape of the nail. The color and texture of the thumb finger nail plate has considered for our research work,



Beau's lines
Lunulae

Vertical lines

White patches



Black Nail
colored

White Nail

Greenish

Gray

Fig-1 Nail Abnormalities

There are eighteen types of abnormalities are possible in finger nails. The types of abnormalities are considered for the research work and some of the abnormalities are listed as follows:

- (1) **Horizontal ridges or Beau's Lines:** Injury, infection, mal-nutrition;
- (2) **Infected nails (red, tender, swollen, pus):** Bacterial or yeast infection;
- (3) **Overlarge moons (Lunula):** Overactive thyroid, genetics, self-induced trauma (habit tick). **No Lunula:** Underactive thyroid, genetics;
- (4) **Colorless-White nails:** May indicate anemia
- (5) **Red or dark pink:** poor peripheral circulation;
- (6) **Bluish nails:** Blood may not be receiving adequate oxygen due to respiratory disorders, cardiovascular problems;
- (7) **Yellowish:** fungus, diabetes, psoriasis, use of tetracycline, or heredity;
- (8) **Half white / half pink:** fungal infection or kidney disease;
- (9) **Small white patches:** a sign of injury to the nail matrix, vitamins and minerals deficiencies;
- (10) **Purple or black:** trauma or vitamin B12 deficiency;
- (11) **Brown or**

black streak that begins at the base of the nail and extends to its tip could be a diagnostic clue to a potentially dangerous melanoma.

In this paper, nail image classification has done with the proposed algorithm and compared with the existing algorithms. Section I describe the background truth of the nail plate abnormalities which are used as signs for various diseases and organ failures. Section II narrates how various features and classifiers can be used to extract and classify the diseased areas. Section III elucidates the proposed classifier and Section IV derives the results and discussions.

II. LITERATURE REVIEW

Feature extraction is an important process in image analysis, computer vision and pattern recognition based applications. It is mainly used for dimensionality reduction and to classify with high accuracy. Feature means the mathematical information of an image, which is labelled after the segmentations. The characteristics of an image can be stored into a feature vector (Let say $V=(v_1, v_2, \dots, v_n)$) which is known as Feature extraction. Textures can be defined as spatial variation in pixel intensities (gray values).

The texture features Entropy, Correlation, Homogeneity and contrast are extracted by GLCM (Gray Level Co-occurrence Matrix). The basis of GLCM is assigning the association among two neighboring pixels in single offset as the next order surface. The gray values associations in a result image are altered into the co-occurrence matrix by a specified kernel mask 3×3 , 5×5 , 7×7 and so forth. In the alteration from the image onto the co-occurrence matrix, the neighboring pixels in 0 degree orientation can be used. It contains information of the position of pixels having related gray level values (Dr.K.Ramesh Babu et al, 2013) [6]. Entropy is evaluated of unpredictability that is used to describe the texture or surface of the input image. Its value will be highest when all the elements of the co-occurrence matrix are the similar. The contrast is a measure of strength of a pixel and its neighbor over the image.

Classification is a function that distinguishes feature vectors into different classes. It involves two steps: Learning and Classification. In learning or training step, a classifier is constructed by learning from a training dataset and their relevant class labels.

These predefined classes are determined by a class label attribute which serves as a category or class. In classification step, the classifier is used to predict the class labels for given new data which is known as testing. In general, test data are used to estimate the accuracy of the classification rules. The accuracy is the percentage of test data set that are correctly classified by the classifier.

Then randomly only some images are chosen up which are measured to be the centroid of every cluster. Then the distance (i.e, similarity) among the images is establish out by computing the *Euclidean distance* among the feature vectors of the images being measured. The images are supposed to be parallel if their Euclidean distance is equivalent to the *threshold* value (user defined value). Thus the related images are grouped into a cluster (here, the cluster is a directory). Having placed all the images into their individual clusters, for every cluster based upon the containing images, novel centroid image is found.

Gulnaz Alimjan et al 2017 [4] proposed a hybrid classifier SV-NN to the remote sensing satellite images. The SVM is applied to the learning phase and KNN is applied to the classify the sample. SVM categories the certain classes by minimum average distance method by Euclidean distance method. The hybrid classifier has compared with conventional classifiers such as KNN, SVM and ANN with the performance metrics overall accuracy and kappa coefficient.

III. METHODOLOGY

The proposed work considers the finger nail image classification process of Hybrid classification algorithm is derived in this portion. This paper introduced a good finger nail image classification process of Combining KNN with SVM classification algorithm. Previous to classification, the Transitioned Row-Column sorting (TRCSF) denoise filter, Graph Based Threshhistomatch Segmentation (GTHM) segmentation process and the feature is extracted using GLCM feature method. The overall finger nail classification procedure flow diagram is illustrated in figure 2.

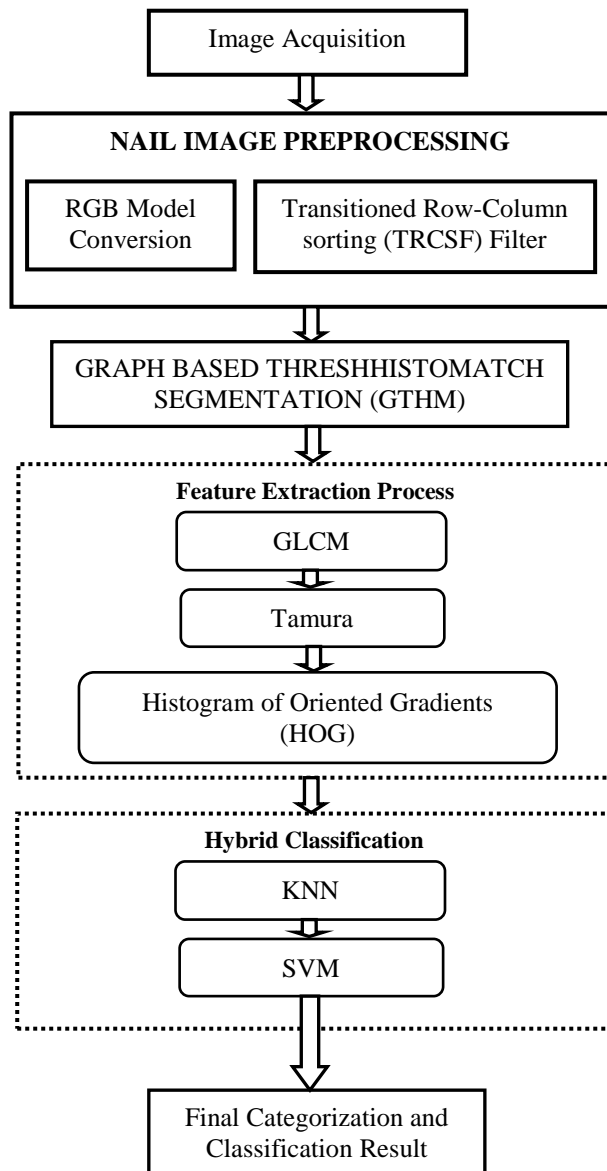


Fig-2: Flow-chart of Proposed work

a) Image Acquisition, Image Preprocessing and GTHM:

Thahira Banu.V, Dr. M. Renuka Devi, 2020 [2-3], already processed the following techniques: Image Acquisition, Transitioned Row-Column sorting (TRCSF) denoise Filter, Graph Based Threshhistomatch Segmentation (GTHM) for Segmentation. The finger nail images may contain noises thus it reduces the redundant noises to notice the injuries. So, TRCSF filter is executed to get improved the PSNR value and to increase the fineness of the image. After TRCSF de-noising process, segmentation method is completed with GTHM Algorithm can be done with skin detection and threshhistomatch algorithm is efficiently using the global and automatic thresholding techniques which enhance the step convolution of the proposed system which segments the disease portions from the

denoised image results that are shown in Fig 3. The above three procedures are done before extracting the features of the finger nail images.

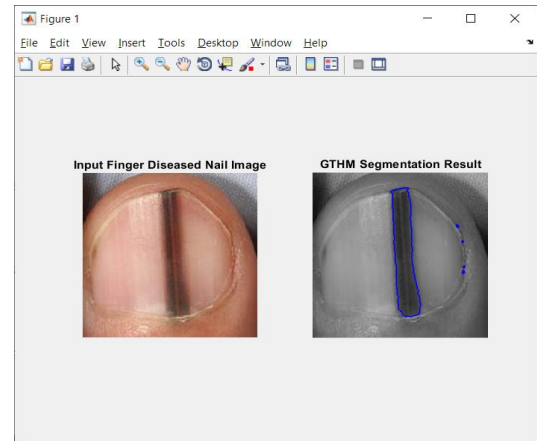


Fig-3: GTHM Segmentation Result

b) Feature Extraction

After the GTHM segmentation procedure a Gray Level Co-occurrence Matrix (GLCM) Feature based Extraction with contrast, correlation, energy, homogeneity, mean, standard deviation, entropy, RMSE, variance, smoothness, kurtosis, skewness, ID (Image Difference), HOG (Histogram Oriented Gradient) and Tamuraf features are executed. HOG feature descriptor focuses on the disease structure or the disease shape of an object. Tamura features will execute on homogeneous texture images. This paper proposed GLCM based Feature Extraction algorithm to obtain the trained features which will be used for Hybrid Classification procedure. The feature extraction results are shown in Figure 4.

	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18
24	0.0047	0.0029	0.0029	0.0017	0.0017	0.0029	0.0029	0.0017	0.0017	0.0047	0.0047	0.0029	0.0029	0.0029	0.0047	0.0029	0.0029	0.0047
25	0.0074	0.0036	0.0036	0.0023	0.0023	0.0036	0.0036	0.0023	0.0023	0.0074	0.0074	0.0036	0.0036	0.0036	0.0074	0.0036	0.0036	0.0074
26	0.0163	0.0069	0.0069	0.0045	0.0045	0.0069	0.0069	0.0045	0.0045	0.0163	0.0163	0.0069	0.0069	0.0069	0.0163	0.0069	0.0069	0.0163
27	0.0308	0.0136	0.0136	0.0091	0.0091	0.0136	0.0136	0.0091	0.0091	0.0308	0.0308	0.0136	0.0136	0.0136	0.0308	0.0136	0.0136	0.0308
28	0.0525	0.0213	0.0213	0.0141	0.0141	0.0213	0.0213	0.0141	0.0141	0.0525	0.0525	0.0213	0.0213	0.0213	0.0525	0.0213	0.0213	0.0525
29	0.0816	0.0337	0.0337	0.0225	0.0225	0.0337	0.0337	0.0225	0.0225	0.0816	0.0816	0.0337	0.0337	0.0337	0.0816	0.0337	0.0337	0.0816
30	0.1046	0.0432	0.0432	0.0287	0.0287	0.0432	0.0432	0.0287	0.0287	0.1046	0.1046	0.0432	0.0432	0.0432	0.1046	0.0432	0.0432	0.1046
31	0.1322	0.0545	0.0545	0.0361	0.0361	0.0545	0.0545	0.0361	0.0361	0.1322	0.1322	0.0545	0.0545	0.0545	0.1322	0.0545	0.0545	0.1322
32	0.1648	0.0676	0.0676	0.0446	0.0446	0.0676	0.0676	0.0446	0.0446	0.1648	0.1648	0.0676	0.0676	0.0676	0.1648	0.0676	0.0676	0.1648
33	0.2025	0.0827	0.0827	0.0549	0.0549	0.0827	0.0827	0.0549	0.0549	0.2025	0.2025	0.0827	0.0827	0.0827	0.2025	0.0827	0.0827	0.2025
34	0.2454	0.0998	0.0998	0.0673	0.0673	0.0998	0.0998	0.0673	0.0673	0.2454	0.2454	0.0998	0.0998	0.0998	0.2454	0.0998	0.0998	0.2454
35	0.2935	0.1189	0.1189	0.0818	0.0818	0.1189	0.1189	0.0818	0.0818	0.2935	0.2935	0.1189	0.1189	0.1189	0.2935	0.1189	0.1189	0.2935
36	0.3466	0.1400	0.1400	0.0975	0.0975	0.1400	0.1400	0.0975	0.0975	0.3466	0.3466	0.1400	0.1400	0.1400	0.3466	0.1400	0.1400	0.3466
37	0.4047	0.1641	0.1641	0.1154	0.1154	0.1641	0.1641	0.1154	0.1154	0.4047	0.4047	0.1641	0.1641	0.1641	0.4047	0.1641	0.1641	0.4047
38	0.4678	0.1912	0.1912	0.1373	0.1373	0.1912	0.1912	0.1373	0.1373	0.4678	0.4678	0.1912	0.1912	0.1912	0.4678	0.1912	0.1912	0.4678
39	0.5359	0.2213	0.2213	0.1626	0.1626	0.2213	0.2213	0.1626	0.1626	0.5359	0.5359	0.2213	0.2213	0.2213	0.5359	0.2213	0.2213	0.5359
40	0.6090	0.2544	0.2544	0.1913	0.1913	0.2544	0.2544	0.1913	0.1913	0.6090	0.6090	0.2544	0.2544	0.2544	0.6090	0.2544	0.2544	0.6090
41	0.6881	0.2905	0.2905	0.2234	0.2234	0.2905	0.2905	0.2234	0.2234	0.6881	0.6881	0.2905	0.2905	0.2905	0.6881	0.2905	0.2905	0.6881
42	0.7732	0.3296	0.3296	0.2595	0.2595	0.3296	0.3296	0.2595	0.2595	0.7732	0.7732	0.3296	0.3296	0.3296	0.7732	0.3296	0.3296	0.7732
43	0.8643	0.3727	0.3727	0.3006	0.3006	0.3727	0.3727	0.3006	0.3006	0.8643	0.8643	0.3727	0.3727	0.3727	0.8643	0.3727	0.3727	0.8643
44	0.9614	0.4198	0.4198	0.3567	0.3567	0.4198	0.4198	0.3567	0.3567	0.9614	0.9614	0.4198	0.4198	0.4198	0.9614	0.4198	0.4198	0.9614
45	1.0645	0.4709	0.4709	0.4278	0.4278	0.4709	0.4709	0.4278	0.4278	1.0645	1.0645	0.4709	0.4709	0.4709	1.0645	0.4709	0.4709	1.0645

Fig-4: Feature Extraction Result

c) Hybrid Classification

This paper presents a new method of Hybrid classification algorithm is an extension of K Nearest Neighbor (KNN) and Support Vector Machine (SVM) Classification. A SVM (Support Vector Machine) method works on the principle of structural risk minimization in order to find the best hyper plane that separates 10 classes (1 for Beau's lines, 2-Black Nails, 3-Dark reddish,4-greenish, 5-half-white nail, 6-lunulae-7-pinkish,8-vertical line, 9-white nails and 10-white patches). The SVM is executed on the training image feature instances to attain the reduced Support Vectors (SVs) for every of the sample classes. To combine SVM with KNN classification model used for training and testing classification purpose. Meanwhile, this paper presents categorization process done with trained and testing feature with help of Euclidean distance method with appropriate threshold value.

$$dist(P, Q) = \sqrt{\frac{\sum_{m=1}^N (P_m - Q_m)^2}{std(Q_m)}} \text{ eqn. (1)}$$

Where, P and Q are trained and testing feature vectors. After the SVM process, a K Nearest neighbor classifier (KNN) is used to classify a testing feature, i.e. the mean Euclidean distance among the testing image feature data point to every set of SVs from different groups is calculated and the category with least distance. A finger nail image GLCM trained feature extraction dataset is given; it chooses the k nearest samples from the classified training data and determines the class considering the most representative samples. The choice of the parameter k ($k \in N$) is determined by the user (i.e., $k = 4$ is defined), this option depends on the image data. A kNN search on the local features of the training set (T_s) is executed. The consequence of such operation is a list of labeled features l_i belonging to T_s ordered with respect to decreasing values of the similarity $sim(l_x, l_i)$. The class label $cl^k(l_x)$ assigned to the trained data by the classifier is the class $c_j \in C$ that maximizes the sum of the similarity between l_x and the features l_i , labeled c_j , in the kNN results list $k(l_x)$. Formally, the hybrid classification have to compute a score $sc^k(l_x, c_j)$ for each class,

$$sc^k(l_x, c_j) = \sum_{l_i \in k(l_x): cl(l_i) = c_j} sim(l_x - l_i) \text{ eqn. (2)}$$

Then the predicted label cl^k and the confidence v are defined as follows:

$$\begin{cases} cl^k(l_x) = \arg \max_{c_j \in C} sc^k(l_x, c_j) \\ v(cl^k, l_x) = 1 - \frac{\arg \max_{c_j \in C - cl^k(l_x)} sc^k(l_x, c_j)}{\arg \max_{c_i \in C} sc^k(l_x, c_i)} \end{cases} \text{ eqn. (3)}$$

Pseudo Code: Hybrid Classification

Input: $GTHM$ image, Trained feature Tr_f , Test feature Ts_f , Class label cl

Output: Classification Result Class cl (1 for Beau's lines, 2-Black Nails, 3-Dark reddish,4-greenish, 5-half-white nail, 6-lunulae-7-pinkish,8-vertical line, 9-white nails and 10-white patches)

Initialization

Step 1: Choose k value.

Step 2: Calculate distance value of trained and test finger nail feature

Step 3: Extract the disease category image feature with appropriate threshold.

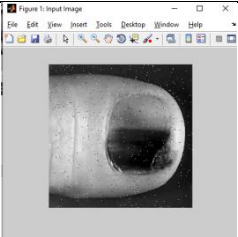
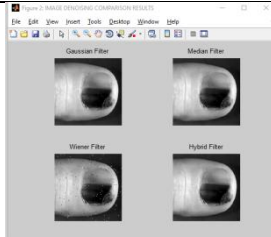
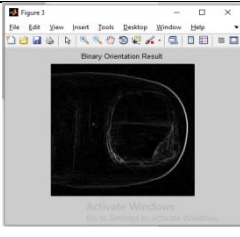
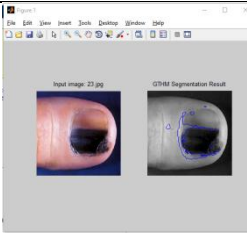
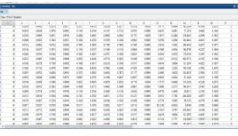
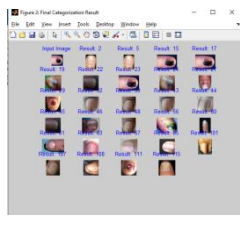
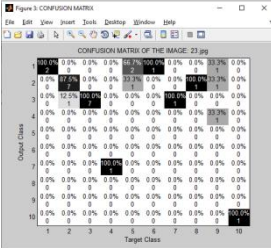
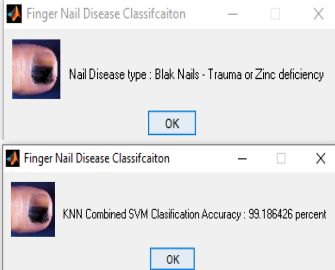
Step 4: Calculate disease similarity score using eqn, (2).

Step 5: Calculate predicted disease class label using eqn. (3)

Step 6: Finger Nail disease classification Result

IV. EXPERIMENTAL RESULT

The experimental result has been calculated the performance of Hybrid Classification algorithm. The results implemented on Intel I5-6500 series 3.20 GHz 4 core processor, 8GB main memory, and runs on Windows 10 operating system, on which MATALB R2018a. In this experimental work is carried for 480 real time thumb finger nail images which are capture in cannon D50 camera and the overall classification accuracy for all 480 images are calculated with image database could be applied to proposed Hybrid Classification with existing KNN, SVM algorithms which can be flexibly configured to forecast the classification accuracy of data base to meet the needs of various test requirements.

Input Image	Noise removed and Enhanced image by TCRS Filter	Binary orientation result	Segmentation
[1]	[2]	[3]	[4]
			
Feature Extraction	Categorization	Confusion matrix	Result & Accuracy
[5]	[6]	[7]	[8]
			

V. PERFORMANCE COMPARISON

Table 1 shows the evaluation of real time finger nail disease database of sample five images of accuracy values for various proposed Hybrid classification with existing KNN and SVM classification algorithms.



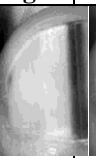
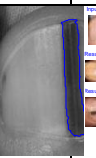

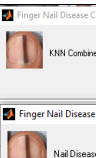
Original Image	Noise added	Pre-processing	Segmentation	Categorization	Result
					
Fig - 95.jpg	Fig - Salt & pepper noise	Fig - Result of TCRS Filter	Fig - Result of GTHMS	Fig - Result of Feature extraction	Fig - Result of KNNwith SVM classifier

IMAGE NAME	SVM	KNN	PROPOSED METHOD
1.jpg	57.894737	78.847368	99.141886
2.jpg	79.166667	79.166667	99.421761
3.jpg	11.764706	73.529412	99.915736
4.jpg	90.000000	85.714286	97.792007
5.jpg	79.166667	79.166667	99.959492
MEAN ACCURACY	63.590555	79.304980	99.246217

Fig 5: Overall Steps Carried Out By this hybrid classification with Sample Input Image

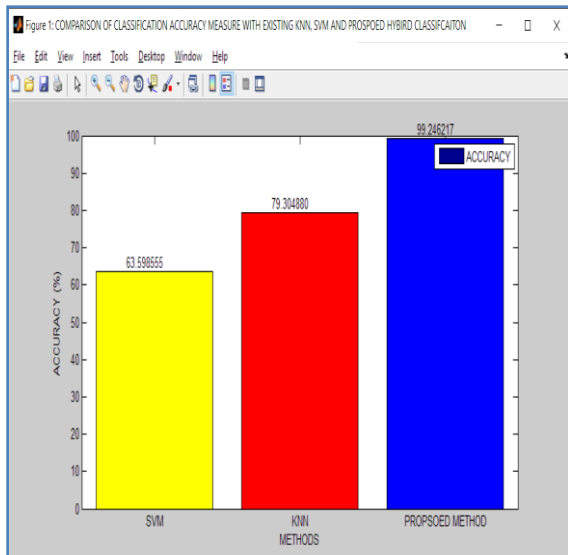


Fig.6: Comparison of Classification Accuracy measures of sample five finger nail images

Table 2 shows the evaluation of Finger nail disease classification measures of accuracy on nail image database. This work is carried for 480 images which are available in finger nail database and the overall accuracy for all 480 images is calculated. The average Classification accuracy of 480 finger nail images and the ratio is defined by,

$$\text{OverAll Accuracy} = \text{abs}(\text{mean}(\text{Classifcaiton_Accuracy})) \text{ eqn. (4)}$$

Table 2: Comparison of Overall Accuracy of thumb finger nails 480 Images

MEASURE	SVM	KNN	HBRID CLASSIFICATION
ACCURACY	71.1066	80.2592	98.5541

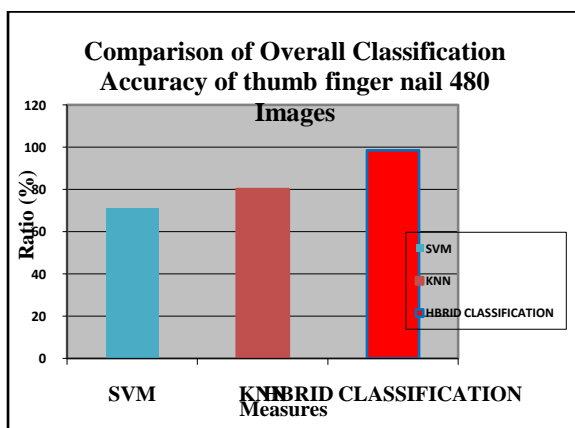


Fig.6: Chart of Classification Accuracy measures of 480 thumb nail images

IV. CONCLUSION

In this paper, thumb finger nail image hybrid classification methods are examined and the results are shown. Nail abnormalities can be manually evaluated but could not exactly differentiate the colors and textures by all because of the naked eyes. An automated system can ensure error free identification of the abnormalities at lesser time. This paper has proposed a new classifier to find the abnormalities based on the various features that are extracted from the source image. The proposed system uses machine learning algorithms and image processing techniques to segment and classify the images. To overcome the drawbacks of the existing algorithms, TCRS filter, GTHMS-Graph Thresh Histo Match Segmentation and hybrid KNNwithSVM classifier algorithms are used. The hybrid classifier classifies the nail plate abnormalities very efficiently and provides better accuracy than SVM and KNN classifier. The experimental results shows the performance metrics such as classification accuracy with the existing SVM algorithm obtains 71.1066%, KNN algorithm obtains 80.2592% whereas proposed Hybrid classification yields 98.5541% overall accuracy.

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