

HMM Classification Technique for Breast Cancer Detection

Harpinder Singh¹, Amandeep kaur², Dr. G. N Verma³

¹Research Scholar, ²Assistant Professor, ³Principal

¹²³Sri Sukhmani Institute of Engineering & Technology, Dera Bassi, Punjab

Abstract- In the existing paper, weight based algorithm is used to classify the normal and cancer cells and it is been analyzed that weight based algorithm taken long time to classify the data. To classify the data in minimum amount of time HMM classifier is used for classification. The second issue with weight based algorithm is of accuracy. As due to weight calculation accuracy of classification is less which can be improved with the use of Bayesian classifier. In the feature selection part on three features are used which are mass, density and margin. In the improvement more features like tissue color will be added which improve detection rate. The simulation is performed in MATLAB and it is been analyzed that proposed technique performs well in terms of fault detection rate, accuracy, MSE and PSNR.

Keywords- MSE, PSNR, Bayesian classifier, HMM classifier

I. INTRODUCTION

Image processing is computer imaging where application includes an individual in the visual circle. At the end of the day the image are to be analyzed and a followed up on by individuals. There are two sorts of strategies utilized for image processing to be specific, simple and digital image processing. Simple image processing can be utilized for the printed versions like printouts and photographs. To distinguish text from the natural images scenes inverse to faxes, business cards and sweep of printed pages is essential stride of computer for computer vision applications [1]. Along these lines, the computerized help for outwardly hindered, automatic decoding of businesses and robotic navigation in urban situations is essential [2].

The extraction of image content description and their associated matching is called feature detection. The main step of which is required is memory consuming and redundant raw images. With the late upheaval of multimedia-empowered frameworks, the requirement for multimedia retrieval has expanded by leaps and bounds. Because of the multifaceted nature of multimedia contents, image comprehension is a troublesome albeit-fascinating theme of exploration, inside the domain of multimedia retrieval. Extracting profitable learning from a substantial scale multimedia store, normally alluded to as "multimedia mining", has recently caught up for lost time as a domain of enthusiasm amongst specialists. The main techniques of feature extractions are Sobel Edge Detection, Canny Edge Detection and Hough Transformation [3].

Breast cancer is a malignant tumor that begins in the cells of the breast. A malignant tumor is a gathering of cancer cells that can develop into (attack) close-by tissues or spread (metastasize) to removed parts of the body. Breast cancer happens for the most part in ladies, yet men can get it, too. Most breast cancers start in the cells that line the ducts (ductal cancers). Some start in the cells that line the lobules (lobular cancers), while somewhat number start in various tissues. Certain techniques are utilized for the detection of breast cancer in women at early age or in the later stage. There are different factors that influence the reason for it furthermore there are various techniques proposed for its detection and removal. Here, there are certain image processing strategies that have been utilized as a part of distinguishing the breast cancer [4].

The principle goal of image segmentation is to extricate different features of the images which can be merged or split keeping in mind the end goal to construct objects of enthusiasm on which analysis and interpretation can be performed. To enhance visibility of the abnormalities to identify breast cancer in mammograms to assist examiners and also automatic breast cancer detection systems, mammogram contrast should be enhanced. Removal of noise is the key for enhancement of contrast of an image, particularly for mammograms the micro-calcification size is near noises. Noise ought to be lessened though micro-calcification should be enhanced [5]. Mammogram image enhancement is the way toward controlling mammogram images to expand their contrast and lessen the noise present to encourage radiologists in the detection of abnormalities. Mammogram is the essential test to distinguish breast cancer. Amid the mammogram test, the Iron Radiation that goes into the breast indicates inside parts of the body furthermore the suspicious locale. It indicates tissues of breast and veins. In the wake of finishing the mammogram test, the outcome will be appeared in X-beam film sheet. Early detection and appropriate medicinal registration are mandatory however in the meantime legitimate diet and sustenance additionally battles with malignant cancer cells [6].

II. LITERATURE REVIEW

Xiong Wang, et.al in paper [7], proposed the feasibility of Contrast-Enhanced Thermoacoustic Imaging (CETAI) for breast cancer detection is researched by a methodical computational study utilizing realistic numerical breast phantoms with tumors. Single-walled carbon nanotubes with a

nontoxic concentration are connected as the contrast agents to build the dielectric properties of the breast tumors and upgrade their perceptibility. Complete CETAI models are created and understood for produced thermoacoustic signals by numerical techniques. Back-projection imaging and differential imaging are performed to envision the tumors. The reproduction result is confirmed by another free numerical technique and a preliminary analysis is performed to exhibit the significant point of the CETAI methodology.

Jeff M. Sill, et.al in paper [8], proposed that the Microwave breast cancer detection is based on contrasts in electrical properties amongst sound and malignant tissues. Tissue sensing versatile radar (TSAR) has been proposed as a technique for microwave breast imaging for early tumor detection. TSAR detects all tissues in the volume of interest and adjusts as needs be. Simulation results have demonstrated the feasibility of this framework for recognizing tumors of 4mm in diameter. In this paper, the second generation exploratory framework for TSAR is introduced. Materials with electrical properties like those in the breast are utilized for the breast model. Tumor detection with the trial framework is assessed and contrasted with simulation results.

Maciej Klemm, et.al in this paper [9], proposed an Ultra Wide Band (UWB) microwave system for breast cancer detection is displayed. The system is based on a novel hemispherical real-opening antenna array, which is utilized in a multi-static radar-based detection system. The radar system is intended to be utilized with realistic three-dimensional (3D) breast phantoms, which have been produced, and with real breast cancer patients amid starting clinical trials. Images are framed utilizing two diverse beam forming algorithms and the performance of these algorithms is firstly looked at through numerical simulation. The paper has introduced and thought about two algorithms: DAS and MAMI beam forming. Numerical simulations demonstrated that with great preprocessing equalization of crude measured signals, MAMI gives better results as far as signal-to-clutter ratio, contrasted and standard DAS.

Konstantinos Kontos et.al proposed in this paper [10], a domain of enthusiasm for data mining applications is the investigation of biomedical data which, in combination with the field of image processing, give exhaustive examination to find hidden patterns or conduct. In the present methodology, genetic algorithms are used trying to decrease the list of capabilities to the informative ones and class awkwardness issues were additionally managed by consolidating a hybrid boosting and genetic sub-sampling approach. As respects to the element extraction approach, the possibility of trainable segmentation is borrowed, utilizing Decision Trees as the base learner. Results demonstrate that the best precision and recall rates are accomplished by utilizing a combination of Adaboost and k-Nearest Neighbor. Based on the utility of this strategy, this paper proposes a system for recognizing cancer that could

help medicinal staff and enhance the exactness of identification of tumor tissues.

Hairong Qi et.al proposed in this paper [11], that the utilization of thermal infrared (TIR) imaging in breast cancer study began as ahead of schedule as 1961. Late advances in image-processing capacity have prodded reestablished enthusiasm for the utilization of infrared breast imaging. This paper gives a review of late accomplishments from these angles. IR radiation possesses the area between the obvious and microwave ranges of the range. All objects in the universe emanate radiation in the IR district as a component of their temperature. As an object gets more blazing, it emits more serious infrared radiation, and it radiates at a shorter wavelength. The objective was to demonstrate that because of the advances in image-processing techniques, the pathophysiological-based understanding of thermograms, and the new-generation IR sensing technology, IR imaging is mature for use as a first-line supplement to both wellbeing monitoring and clinical diagnosis.

V. Vishrutha et.al proposed in this paper [12], that breast cancer is a standout amongst the most well-known cancers among ladies. Around two out of three invasive breast cancers are found in ladies with age 55 or more seasoned. A Mammogram (low energy X-ray of breast) done to recognize breast cancer in the early stage when it is impractical feel a lump in the breast. This paper has proposed a technique to recognize microcalcifications and outlined masses and likewise classify them as Benign or malignant. The proposed technique comprises of three steps: The initial step is to discover region of interest (ROI). The second step is wavelet and texture feature extraction of ROI. The third step is classification of identified abnormality as generous or malignant utilizing Support vector machine (SVM) classifier. The proposed strategy was assessed utilizing Mini Mammographic Image Analysis Society (Mini-MIAS) dataset. The proposed strategy has accomplished 92% exactness.

III. PROPOSED WORK

In the proposed technique, the MIRE images are taken and each MIRE image has its unique identification. The proposed system will match various features of the cancer cells to declare that whether note is cancer or not.

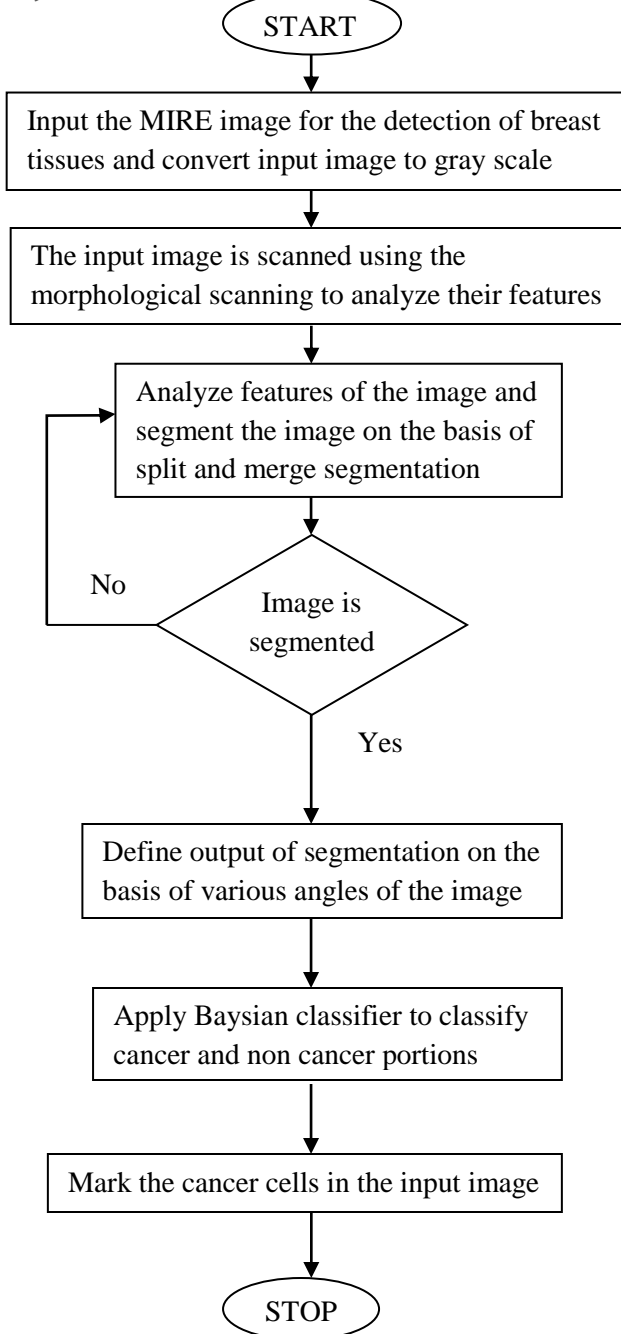


Fig.1: Flowchart of proposed technique

These features are of shape, color and intensity of color breast tissue cells. All these features are extracted using SWIFT and features are matched using nearest neighbour classifier. In this improvement will be proposed in framework in which neighbour classifier will be replaced with Bayesian classifier. This will leads to improve accuracy of cancer detection and improvement in SIFT algorithm will proposed to match

internal features of the currency. In this work, existing base paper technique is implemented for cancer reorganization. In this work, the image is taken as input and that input image will be converted into gray scale.

Steps of Flowchart

1. **Morphological Scanning:** Mathematical Morphology is a strategy for quantitative analysis of spatial structures that goes for analyzing shapes and forms of an object. Mathematical morphology depends on set theory. The shapes of objects in a binary image are spoken to by object membership sets. Objects are associated regions of pixels with value 1, the background pixels have value 0.
2. **Spilt and merge segmentation:** Splitting and merging endeavors to separate an image into uniform regions. The essential representational structure is pyramidal, i.e. a square region of size m by m at one level of a pyramid has 4 sub-regions of size $m/2$ by $m/2$ underneath it in the pyramid.
3. **Bayesian classifier:** A Bayesian classifier depends on the possibility that the part of a (natural) class is to predict the values of features for members of that class. Illustrations are assembled in classes since they have normal values for the features. Such classes are regularly called natural sorts. In a Bayesian classifier, the learning agent manufactures a probabilistic model of the features and uses that model to predict the classification of another case.

IV. EXPERIMENTAL RESULTS

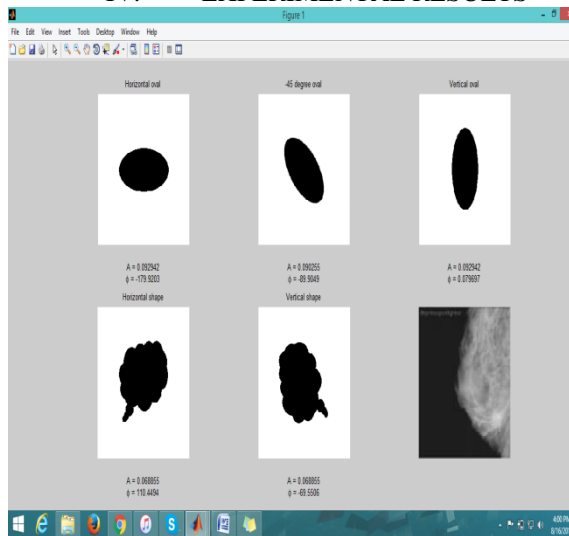


Fig.1: Detection of cancer tissues

As shown in figure 1, the image which is taken as input is the gray scale image. The gray scale is segmented and on various angles its segmentation is done. The segmented image is given as input to Bayesian Classifier.

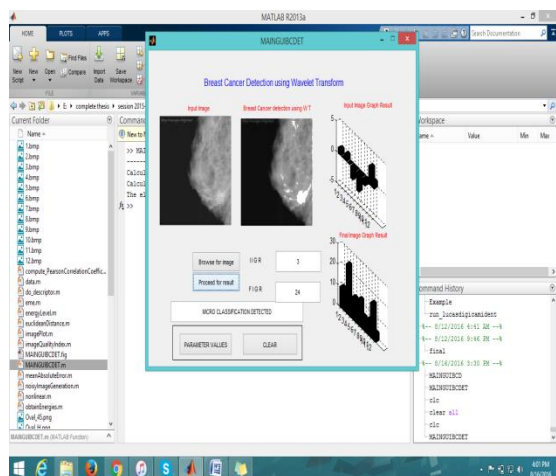


Fig.2: Detection of cancer cells

As shown in figure 2, the image which is taken as input is the gray scale image. The gray scale is segmented and on various angles its segmentation is done. The segmented image is given as input to Bayesian classifier. The Bayesian Classifier will mark the cancer portion in the image.

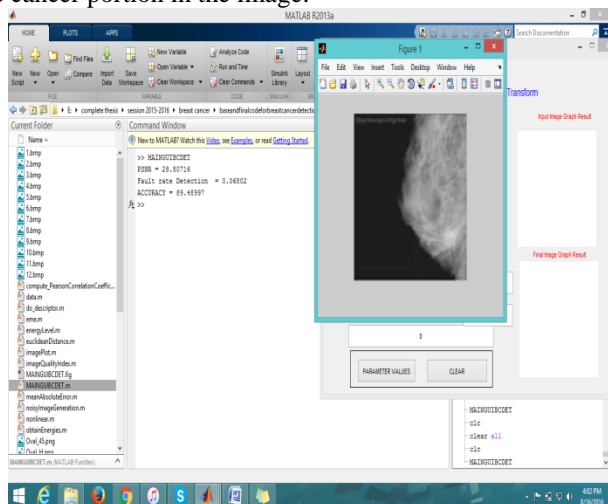


Fig.3: Parameter values

As shown in figure 3, the various parameters are used to analysis the performance of proposed techniques. These parameters are fault detection rate, accuracy, PSNR and MSE values.

V. CONCLUSION

In this work, it is been concluded that to detect breast cancer various techniques has been proposed in the previous times. The most efficient technique of breast cancer detection is based on morphological scanning, split and merge segmentation and on nearest neighbor classifier. In this work, to improve efficiency of the breast cancer detection nearest neighbor classifier is replaced with Bayesian classifier. The

split and merge segmentation will split the input image on the basis of their properties. The output of split and merge segmentation is given as input to Bayesian classifier which will classify the features of the basis of their properties. The cancer and non cancer cells are marked with different colors. The simulation is performed in MATLAB and it is been analyzed that proposed technique performs well in terms of fault detection rate and accuracy.

VI. REFERENCES

- [1]. Rangaraj M. Rangayyan, Liang Shen, Yiping Shen, J. E. Leo Desautels, Heather Bryant, Timothy J. Terry, Nataalka Horeczko, and M. Sarah Rose, "Improvement of Sensitivity of Breast Cancer Diagnosis with Adaptive Neighborhood Contrast Enhancement of Mammograms", 1997, IEEE TRANSACTIONS ON INFORMATION TECHNOLOGY IN BIOMEDICINE, VOL. 1, NO. 3
- [2]. Aziz Makandar, Bhagirathi Halalli, "Breast Cancer Image Enhancement using Median Filter and CLAHE", 2015, International Journal of Scientific & Engineering Research, Volume 6, Issue 4
- [3]. Martin O Halloran, Edward Jones, and Martin Glavin, "Quasi-Multistatic MIST Beamforming for the Early Detection of Breast Cancer", 2010, IEEE TRANSACTIONS ON BIOMEDICAL ENGINEERING, Vol. 57, No. 4
- [4]. R. Guzmán-Cabrera, J. R. Guzmán-Sepúlveda, M. Torres-Cisneros, D. A. May-Arrijoa, J. Ruiz-Pinales, O. G. Ibarra-Manzano, G. Aviña-Cervantes, A. González Parada, "Digital Image Processing Technique for Breast Cancer Detection", 2013, Springer, 34:1519-1531
- [5]. Anuj Kumar Singh and Bhupendra Gupta, "A Novel Approach for Breast Cancer Detection and Segmentation in a Mammogram", 2015, Procedia Computer Science 54 676 – 682
- [6]. D.SURYA GOWRI, Dr. T. AMUDHA, "A Review on Mammogram Image Enhancement Techniques for Breast Cancer Detection", 2014, IEEE, 978-1-4799-3966-4/14
- [7]. Xiong Wang, Tao Qin, Russell S. Witte, and Hao Xin, "Computational Feasibility Study of Contrast-Enhanced Thermoacoustic Imaging for Breast Cancer Detection Using Realistic Numerical Breast Phantoms", 2015, IEEE TRANSACTIONS ON MICROWAVE THEORY AND TECHNIQUES, VOL. 63, NO. 5
- [8]. Jeff M. Sill, and Elise C. Fear, "Tissue Sensing Adaptive Radar for Breast Cancer Detection—Experimental Investigation of Simple Tumor Models", 2005, IEEE TRANSACTIONS ON MICROWAVE THEORY AND TECHNIQUES, Vol. 53, No. 11
- [9]. Maciej Klemm, Ian J. Craddock, Jack A. Leendertz, Alan Preece, and Ralph Benjamin, "Radar-Based Breast Cancer Detection Using a Hemispherical Antenna Array—Experimental Results", 2009, IEEE TRANSACTIONS ON ANTENNAS AND PROPAGATION, Vol. 57, No. 6
- [10]. Konstantinos Kontos and Manolis Maragoudakis, "Breast Cancer Detection in Mammogram Medical Images with Data Mining Techniques", 2013, AIAI IFIP AICT 412, pp. 336-347

- [11].Hairong Qi and Nicholas A. Diakides,” Thermal Infrared Imaging in Early Breast Cancer Detection”, 2009, Advances in Pattern Recognition, 10.1007/978-1, pp 84800-277-7 6
- [12].V. Vishrutha and M. Ravishankar,” Early Detection and Classification of Breast Cancer”, 2014, (FICTA) Vol. 1, Advances in Intelligent Systems and Computing 327