



# A Survey of Breast Cancer Detection and Classification Based on Texture Feature

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**Abstract:** Breast cancer is second most dangerous disease in the world after the lung cancer among women. Because of this reason, breast cancer detection is most focused area by many researchers. Digital mammographic images and advanced techniques are required by many researchers for developing Computer aided detection (CAD) for breast cancer detection. This study describes the recent advances in image processing and machine learning techniques for breast cancer detection. The study shows that Local Binary Pattern method used for feature extraction and Support vector machine for classification as foremost technique used for breast cancer detection. The comparative study of literature work summarizes the effectiveness of different approach used by researchers for breast cancer detection.

**Keywords:** breast cancer detection, CAD, image segmentation, feature extraction etc.

## I. INTRODUCTION

Breast cancer is common and one of the deadly disease found in women. Prevention of the breast cancer in its early stage is difficult as the cause of the disease is still not known. X-ray mammography is very useful technique for screening and early stage detection of cancer. Regular mammography can help discover cancer in its beginning stage, when treatment is best. A mammogram can discover breast changes that could be disease years before physical side effect develop. Previous research shows that women who have regular mammogram can benefits by discovering breast cancer at an early stage. Such women don't need treatments like surgery to remove the breast, which also help them to cure the cancer.

Women who are at high risk for breast cancer based on certain factors should get a Magnetic resonance Imaging (MRI) and a mammogram every year. The American cancer society recommends against MRI screening for women whose lifetime risk of breast cancer is less than 15%. If MRI is used, it should be in addition to, not instead of a screening mammogram. This is because although an MRI is more likely to detect cancer than mammogram, it may still miss some cancer that a mammogram would detect [1].

Identifying masses asymmetries are difficult for radiologist due to inexperience, they may give wrong output because the mammograms are very sensitive to analyse. Computer Aided Detection (CAD) technology can improve the performance of radiologist. The CAD technique can minimize the events of mistakes conferred by the specialists. CAD system can characterize the normal and abnormal, benign and malignant case from digital mammogram by extracting the relevant features.

In computer aided detection segmentation of the region of interest that is breast profile and pictorial muscle is very important preprocessing stage [3]. The pictorial muscle in mammogram biases detection procedure; therefore it is essential to remove the pictorial muscle during preprocessing. Jawad Nagi et al. [3] present an automated technique for mammogram segmentation. The proposed algorithm make the use of morphological preprocessing and seeded region growing (SRG) algorithm to remove digitization noise, suppress radiopaque artifacts, separate background region from the breast profile region and remove the pictorial muscle, for highlighting the breast profile region.

This survey focuses on different schemes used for breast cancer detection and their comparison based on the result obtained through respective research. The Section II describes different methods for development of algorithms used for breast cancer detection, section III evaluates the methodology for development of CAD system and section IV summarizes the study.



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## II. RELATED WORK

R. Guzman Caberrera et al. [2] propose a computer Aided Detection scheme to separate suspicious region containing the masses from background parenchyma using local entropy. The proposed algorithm composed of background and objects texture segmentation and extraction. The texture features of the background and objects are obtained by using local entropy. In the proposed work, window size of 9\*9 was used to compute local entropy, after that the properties of the local textures are extracted based on the adjacent neighbours of current pixel. The regions of interests are identified and then extracted using both the original input image and the background texture as a contrast mask. The experiment is evaluated on the mammographic images from DDSM database. However, the study proposed is highly dependent on an adequate value of the reference gray level to achieve a successful segmentation and extraction of the suspicious region.

Pavel kral et al. [4] proposes a method for breast cancer detection at an early stage, using LBP features. The background image and breast image is separated using Otsu method. The Local binary patterns values for breast image are calculated, after which image is divided into a set of the square cells lying on a regular grid. Feature vectors are computed for each cell as a histogram of the LBP values. Every cell is then represented by one feature vector. Support vector machine (SVM) based on uniform LBP histogram is used to classify, whether the corresponding region is normal or cancerous carried out on mammographic images obtained from DDSM and MIAS database. The proposed study achieves the accuracy approximately 84%.

Leonardo de oliveira et al. [5], proposes a method for detection of masses in mammographic images using k-means and support vector machine. Image segmentation is done using k-means algorithm and texture characteristics of the segmented area is obtained through co-occurrence matrix and shape descriptor. Based on the texture characteristics of the segmented region, support vector machine classifies the normal and cancerous tissues. The proposed algorithm is evaluated on the mammographic images extracted from Digital Database for Screening mammography. There are certain limitations over the use of k-means algorithm used in the study, which can't detect the masses due to lack of ability to differentiate between mass region and neighbouring region (dense breast tissue). The accuracy achieved by these work is over 85%.

Anuj kumar et al. [6] proposes noteworthy research which composed of detection and segmentation of region of malignant tissue. The method detects the cancerous area, then segment the respective area. In a detection stage, by applying averaging filter and thresholding operation to an original input image, cancerous region is detected. With the help of rectangular window over the detected cancerous region, max-mean and least variance techniques are applied to find the malignant tissues. Finally, by applying morphological closing operation, tumour patch is detected and the region boundary is detected using image gradient technique. As per authors claim, the proposed algorithm is simple and fast because of the basic image processing technique and successful in segmentation of the malignant tissues of the mammographic images. However, there is a dependency on manual selection of threshold parameter and size of the averaging filter.

S. Naresh et al. [7] proposes an approach for breast cancer detection, which uses completed Local Binary Pattern (CLBP) for texture feature extraction with support vector machine (SVM) for classification. The experiments are performed on MIAS database. In the work, morphological operations are carried out to remove noisy data, enhance and contrast the input image. Author claims that CLBP gives robust result as it consider both sign and magnitude features compare to the LBP, which consider only sign parameter, leads to the loss of other important information.

A significant research of implementing Eigen faces approach for face classification to mass detection is proposed by Arnau Oliver et al. [8]. The author present some similarity between face detection and mass detection algorithm, in which the face from the image is detected and then it is assigned to a person. The eigenfaces approach for mass detection has two limitations, which are the multiple size of RoI's and the variance of the gray-level range. In the present work, a uniform distribution model for equalizing the images to overcome the problem of gray-level range.

The problem of multiple size of RoI's is solved using 3 different ways: 1. collecting the database of RoI's in different group based on their size.

Whenever a new RoI is detected, based on their size, corresponding collection will be referred. 2. by cropping the RoI's by using same size. 3. by cropping the exact size of detected RoI and resize it. With the help of leave-out methodology and ROC analysis, performance of above mentioned approaches was calculated which was different for each one. The best result obtained through this work is  $A_z=0.92$ .

Eanes torres Peirira et al. [9] proposes noteworthy classification method to detect six different type of breast cancer namely: CALC, CIRC, SPIC, MISC, ARCH, ASYM. The proposed work is based on LBP feature and SVM classifier. LBP has two important parameters: - P (Number of neighbours) and R (Radius of comparison). The selection of this parameter gives the less or more number of LBP features. In proposed experiment, initially gave 66,080 LBP features by using  $P= \{4, 8, 16\}$  and  $R= \{1, 2\}$ , which increased the processing time of SVM classifier. This problem is overcome by adjusting the parameter ( $P= 4, R= 1$ ) and ( $P= 8, R= 1$ ), resulting in 277 LBP features, used in the study.



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Experiment was evaluated on images form MIAS database. The performance is measured using ROC, which is greater than 0.8.

Arнау Oliver et al. [10] propose an approach focusing on reduction of false positive mass detection using Local Binary Pattern. A group of 1792 uncommon region of interest was obtained from DDSM database. The RoI images divided into 5\*5 rectangular regions. Each rectangle in 5\*5 rectangle regions is divided into 3\*3 regions and LBP features extracted to improve the quality of final RoI descriptor. Textural descriptors are independently obtained with the help of LBP for each region in 5\*5 rectangular regions and then combined to form a global descriptor of RoI based on textural information. The classical LBP histogram is thus amplified to a spatially enhanced histogram which encodes both the local region and the spatial relation of the mass. SVM with a polynomial kernel is used to differentiate RoIs between a true mass and normal parenchyma. From 1792 extracted RoIs, 256 characterize a true mass and remaining normal but suspicious tissue. The performance of research is evaluated using leave one-out strategy and receiver operation characteristics with value  $A_z = 0.909 \pm 0.043$ .

Spandana et al. [11] presented a notable research for breast cancer detection at an early stage with the help of Matlab and Labview. The extraction of the mass with high intensity value is done by obtaining the mask of the mass by applying manual thresholding using histogram of the image. In Labview, region of mass is extracted by performing logical AND operation to the original image and mask. The border of the mass, which are obtained through morphological operator, are superposed on the original image with the help of logical OR operation in labview. Texture, wavelet and geometric features of the RoI of segmented masses are measured to differentiate between malignant and benign. For experiment evaluation, mammographic images of 14 patients were acquired from KIMS, Hyderabad. Most of the calculated features help in classification of benign and malignant masses.

Table I shows the comparative study of all researches discussed in literature review.

TABLE I COMPARATIVE LITERATURE WORK FOR DETECTION OF BREAST CANCER

Author	Image Database	Feature	Remark
Caberrera et al. [2]	DDSM database	Texture feature	Highly dependent on a reference gray level
Pavelkral et al. [4]	DDSM and MIAS database	LBP features	Accuracy achieved – 84%
Anujkumar et al. [5]	---	Gradient	Dependent on threshold parameter selection and size of the averaging filter
Oliveira et al. [5]	DDSM database	Texture characteristics	Accuracy achieved – 85%
S. Naresh et al. [7]	MIAS database	LBP features	Results of CLBP are better than LBP
Oliver et al. [8]	---	Eigen faces approach	Result is $A_z = 0.92$ .
Peirrrira et al. [9]	MIAS database	LBP features	ROC performance > 0.8
Oliver et al. [10]	DDSM database	LBP features	ROC performance = 0.909 +/-0.043
Spandana et al. [11]	Mammographic images of 14 patients were acquired from KIMS, Hyderabad	Texture, wavelet and geometric features	--

### III.METHODOLOGY

This section describes about the different stages involved in order to achieve the purpose of each research mentioned in the literature review. The different stages re composed of preprocessing of the input image for extracting the features, which are then applied as an input to the classifier. The output of the classifier differentiates the normal, benign and malignant cases from applied digital mammographic images.

#### A. Image database.

Digital mammographic images are easily available on internet which can be downloaded from the respective web address. Digital database for screening mammography (DDSM) is one of the databases available from combined efforts of Massachusetts General Hospital, Sandia National Laboratories and the University of South Florida Computer Science and Engineering Department gives approximately 2,500 case studies [13]. The Mammographic Image Analysis Society (MIAS) MiniMammographic Database is another source of digital mammographic images available easily [12]. The mammographic image analysis society is an organization of UK research group has generated a database of digital mammogram.



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### B. Image preprocessing

The purpose image preprocessing is to remove unwanted noise present in the original input image and improve the quality of fine details present in it. This will make it easy for further processing of image in order to achieve the defined aim. The image preprocessing include image enhancement, removal of noise and breast part extraction. Smoothing image pixel using averaging filter [5], Otsu method [4] to separate background from breast region, morphological operations [6], [11], sharpening [2] have been used for preprocessing on digital mammographic images.

### C. Feature extraction

Once the original image is preprocessed, it is possible to extract the relevant features specified in literature. Feature extraction can be described as extracting significant fine information from given input while rejecting all other data. LBP texture feature [4], [7], [9], [10], texture [2], [5], intensity value [11], eigenfaces approach [8], shape [6] feature were extracted from digital mammograms.

### D. Classification

After extracting the relevant feature, the final stage is to classify the obtained data and assigned it to a particular class. For this purpose, classifier like Support Vector Machine [4], [5], [7], [9], [10], Decision tree [8], KNN [8], thresholding [6] were use

## IV. CONCLUDING REMARK

The literature review show that every research discussed here has some dependency on certain parameter in proposed scheme. It can be a reference gray level, threshold selection or size of average filter, which makes the respective algorithm less efficient to achieve defined purpose. Local Binary Pattern method can extract the fine details from image which can be useful to develop CAD system for breast cancer detection in its early stage. From the literature, it is clear that support vector machine is the best option as a classifier for classification between cancerous and non cancerous cases.

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