

# A Review on Various Approaches of Mammography Using Image Compressions

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## Abstract

Breast Cancer screening remains a subject of extraordinary and enthusiastic level headed discussion. Mammography has for a long time been the pillar of bosom malignancy identification and is the main screening test demonstrated to decrease mortality. Despite the fact that mammography remains the best quality level of bosom growth screening, there is an increase in alertness among the subpopulation of women for whom mammography has decrease its affectability. Mammography has likewise experienced expanded examination for false positives and exorbitant biopsies, which increment radiation dosage, cost furthermore, persistent tension. In light of these difficulties, new innovations for bosom tumor screening have been produced, including low measurements mammography.

## Keywords

Image Compression, Lossless compression, Near Lossless compression

## I. Introduction

Breast Cancer is most analysed malignancy, other than skin disease, among females worldwide[1,2]. It is additionally anticipated that the bosom disease is the chief reason for losses which are happening from long ago[3,4]. Different reviews have exhibited that early identification and proper treatment of bosom cancer may decrease the death rate[5,6].

Mammography can't stop or reduce bosom malignancy however are steady just in recognizing the bosom cancer at early stages to enhance the survival rate[2,6]. Standard screening can be an effective procedure to recognize the early side effects of bosom growth in mammographic pictures [7]. Less availability of recent photos taken by advance technology is the first difficult assignment in finding of bosom malignancy[8,9]. There are two types of outcomes i.e. False-positive and False-negative[10]. False-positive outcomes prompt surgeries with benevolent (noncancerous) conditions. False-negatives let the early stage infection with more confused stage with less rate of survival. Now days, use of computer supported strategies have been analysed and yielded for the examination of advanced mammograms. They go for highlighting to zones of interests like sores, masses, and so on, making them noticeable to the radiologists which are useful in improving the probability of early identification of bosom growth from mammographic pictures.

In this paper the first section describe about the general introduction of bosom cancer identification. In the 2nd section the Computer Aided Diagnosis (CAD) is described how it helps the specialists in identifying the cancer. The 3rd section is about the Literature Survey. The 4<sup>th</sup> section explains about the various techniques used for image compression. Finally, the 5<sup>th</sup> section gives the conclusion about the study.

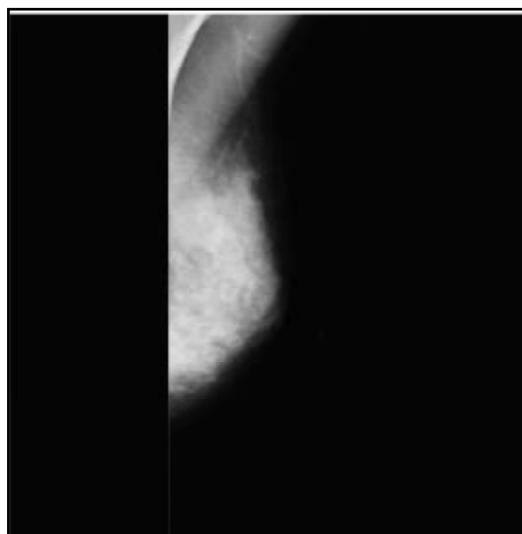


Fig. 1 : A Mammography image of size 1024x1024[11]

## II. Brain Tumor Segmentation

Computer Aided Diagnosis (CAD) is software that helps specialists in the understanding of therapeutic photographs. Imaging processes in X-ray, MRI, and Ultrasound diagnostics give an immense amount of data that the radiotherapist or other expert has to investigate and assess completely in a less duration of time. CAD process new technology photographs for normal appearances and to highlight visible parts which are abnormal, in order to suggest input to support a decision taken by the experts.

Computer Aided Diagnosis (CAD) is generally restricted to check visible structures and parts which are affected by disease. For instance, in mammography CAD highlights micro-calcification clusters and hyper-dense structures in the sensitive tissue. This enables the radiotherapist to do interpretation about the state of the pathology. It officially demonstrates its prosperity not only in decreasing human mistake in reading mammograms and also indicates better and reliable characterization into benevolent and dangerous anomaly.

## III. Literature Survey

M. Pratiwi, Alexander, J. Harefa, and S. Nanda [1] proposed benefits of Radial Basis Function Neural Network (RBFNN) for mammograms characterization in view of Gray-level Co-occurrence Matrix (GLCM) texture based elements. In this

review, usual and unusual bosom pictures are used as standard information taken from Mammographic Image Analysis Society (MIAS) computerized mammogram database. Computer Aided Diagnosis (CAD) help radiologist in grouping different kind of bosom tumor. It officially demonstrated its prosperity not just in diminishing human mistake in interpreting mammograms additionally indicates superior and dependable categorization into usual and unusual anomaly.

S. Mandal and I. Banerjee [2] performed two types of validations to identify tumor: Cross validation and new case testing for over two datasets with unique collaboration of concealed layers and correlating nodes. It was found that, Neural Network (NN) model can order the data with great precision and this will prompt robotized therapeutic analysis system for specific disease. Early detection of disease is based on doctor's ability to understand the type of disease based on their understanding and practice yet a mistake may happen. A variety of treatments has been provided by specialists as of now. Use of different expert techniques for therapeutic analysis of infections in a short time ago becomes worldwide. These rational systems help doctors as a conclusion associate. Presently, various Neural Net, Rough set, Decision Tree, Bayes Network are extremely famous for this purpose.

K. A. M. Junaid [3] exhibited a practical way to categorize the ordinary, cancerous and benignant using two layer semantic net back propagation algorithm. Back propagation algorithm is used to train the semantic net. Parallelization strategies hasten the computing activity and therefore two layers neural net surpass the prior work in terms of precision.

S. Naranje [4] proposed a computerized method using man-made network as decision making apparatus in the field of bosom tumor. Image Processing plays important role in tumor identification when input is in the form of pictures. Feature extraction (numerical parameter) of photo is essential in mammogram categorization. Features are extracted by using picture processing. Different feature extraction techniques are used for categorization of usual and unusual shapes in mammogram. This technique will give maximum exactness at rapid pace.

Kamaldeep Kaur and E. Pooja [5] exhibited the analysis of bosom tumor by using Artificial neural network (ANN) and Support vector machine(SVM). To manage with the distinctive sort of variations from the norm causing tumor, this report comprises of considerable number of modalities which help in recognizing tumor and as well as various techniques of feature extraction. Such modalities can be named as: Mammography, Ultrasound, MRI and so on. At present, Electrical impedence and nuclear drugs are used universally for identification of disease. These modalities depend on the image processing i.e. recognition of deformity take place by scanning and regain knowledge from photographs. But this study depends on mammogram pictures. Before recovering data one should have knowledge about all type of anomalies like: micro classification, masses, structural distortion, asymmetry, bosom density and so on. After the process of extricate the unusual part or can state that ROI (Region of Interest) on which therapy is applied. To remove ROI different strategies are used like region expanding, edge recognition, segmentation and so on. Moreover, feature extraction take place from which a number of features are withdrawn on which feature selection is applied to get higher accuracy.

R. Sehgal and S. Gupta [6] performed characterization in view of semantic network which is more acceptable than other present characterization strategies and concentrate on picture quality and

precision. Picture quality evaluation and improvement are rely on enhancement level where low pre-processing methods are used based on Gabor filter within Gaussian rules; after that the segmentation postulates are implement above the enhanced area of the photo and the input data for feature extraction is attained, additionally depending on the common features, a routinely comparison is made. In this study, the critically analysed features for precise picture comparison are pixel percentage and masking labelling.

R. D. Thakur [7] introduced a relative investigation of PC supported analysis for therapeutic photograph division and edge identification using Semantic Network and Fuzzy Logic. This proposal introduced with case study of skin tumor identification using Artificial Neural Network and Fuzzy Logic. The processed picture is then registered for investigation. The purpose of increasing attention to how Semantic Network and Fuzzy Logic can be applied to these regions will help to discover illness affected region without any mistake can be accomplish using pre-processing and post-processing. Ailment can be recognized in its initial state and can save many lives.

T. Kanimozhi and A. Murthi [8] executed monetary crisis supportive system to process the medicinal pictures. It has been use to examine Melanoma parameters like Asymmetry, Border, Color, Diameter (ABCD) and so on which are computed using MATLAB from skin tumor pictures planning to expand diagnostic algorithms that may enhance triage practices in the crisis sector. Using the ABCD rules for the melanoma skin tumor, we use Semantic Network (SN) in characterization stage with Back Propagation (BP) Algorithm.

M. Joselin, A. Retchal, T. Geetha, and M. M. P. N [9] gives self-executing screening apparatus for real time skin wound by giving the input in the form of photographs that are catches with the help of new technology phones. This system will automatically recognize whether the given photo has melanoma by using texture segmentation, hair identification and exclusion, Gray Level Co-occurrence Matrix (GLCM) for feature extraction and semantic network for classifying strategies. The photo database carry about 50 of skin wound photographs which include Benevolent, Unusual and Melanoma. This awareness will help to find about the drug to seek.

K. Sirinukunwattana, S. E. A. Raza, Y. W. Tsang, D. R. J. Snead, I. A. Cree, and N. M. Rajpoot [10] suggest a novel Neighbouring Ensemble Predictor (NEP) integrated with Convolutional Neural Network (CNN) to more precisely estimate the class logo of recognized cell core. The recommended methodology for identification and categorization does not require division of nucleus and assess them on a huge dataset of colorectal adenocarcinoma pictures, containing more than 20,000 explained notes of nucleus belonging to four distinct classes. Our outcome show that the joint recognition and categorization of the suggested Spatially Constrained-Convolutional Network (SC-CNN) and Neighbouring Ensemble Predictor (NEP) provide the highest average F1 score as compared to other new methodologies. Probably, the suggested techniques could provide benefits to pathology practice in terms of quantitative investigation of tissue components in entire-slide pictures and could possibly lead to a better knowledge of malignancy.

## IV. Image Compression Techniques

### A. JPEG

DCT (Discrete Cosine Transformation)-Based Image Coding Standard .The JPEG/DCT still picture compression has turned into a standard now days. JPEG is planned for compacting full-shading or gray scale pictures of common, real world scenes. To misuse this strategy, a picture is first apportioned into non covered  $8 \times 8$  pieces. A discrete Cosine change (DCT) is connected to each piece to change over the dim levels of pixels in the spatial area into coefficients in the recurrence area. The coefficients are standardized by various scales as indicated by the quantization table given by the JPEG standard directed by some psycho visual confirmation. The quantized coefficients are modified in a request to be further packed by a proficient lossless coding technique, for example, run length encoding, number juggling coding, or Huffman coding. The data misfortune happens just during the time spent coefficient quantization. The JPEG standard characterizes a standard  $8 \times 8$  quantization table for all pictures which may not be proper. To accomplish a superior deciphering nature of different pictures with the same compression by using the DCT approach, a versatile quantization table might be used as opposed to use the standard quantization[12].

### B. Wavelet Transform

Wavelets are capacities characterized over a limited interim and having a normal estimation of zero. The essential thought of the wavelet change is to speak to any discretionary capacity (t) as a superposition of a set of such wavelets or premise capacities. These premise capacities or infant wavelets are acquired from a solitary model wavelet called the mother wavelet, by enlargements or withdrawals (scaling) furthermore, interpretations (shifts). The Discrete Wavelet Transform of a limited length flag  $x(n)$  having N segments, for instance, is communicated by a  $N \times N$  lattice Despite every one of the upsides of JPEG compression plans in view of DCT to be specific straightforwardness, attractive execution, and accessibility of exceptional reason equipment for execution; these are not without their deficiencies. Since the information picture should be blocked, relationship across the piece limits is not dispensed with. This brings about discernible and irritating "blocking curios" especially at low piece rates[12].

### C. VQ Compression

Vector quantization is a strategy from flag handling which permits the demonstrating of likelihood thickness works by the conveyance of model vectors. It works by encoding values from a multidimensional vector space into a limited arrangement of qualities from a discrete subspace of lower measurement. A lower-space vector requires less storage room, so the information is compacted. Because of the thickness coordinating property of vector quantization, the packed information have errors that are contrarily relative to their thickness[13].

### D. Fractal Compression

Fractal compression is a lossy compression technique for computerized pictures, in view of fractals. The technique is most appropriate for surfaces, regular picture, depending on the way that parts of a picture regularly look like different parts of a similar picture. Fractal calculations change over these parts into numerical information called "fractal codes" which are used to reproduce the encoded picture[13].

## E. Lossy Compression techniques

- **Quantization:** Quantization is a process of few to one mapping that change the number of values with just a single value. Scalar and vector quantization are two essential sorts of quantization. SQ (scalar quantization) performs different to one mapping on each value. VQ (vector quantization) replaces each piece of information pixels with the record of a vector in the codebook, which is near the information vector by using a couple of closeness measurements. The decoder just gets each file and looks into the proportional vector in the codebook.
- **Transform Coding :** Transform coding is a typical Strategy for lossy image pressure. It uses a reversible and linear change to decorrelate the real image into an arrangement of coefficients in transform area. The coefficients are then quantized and coded progressively in transform area.
- **Block Transform Coding:** keeping in mind the end goal to abbreviate the calculations, block change coding misuse relationship of the pixels inside various small groups that partition the real image. Accordingly, each square is changed, quantized and coded freely. This technique uses square  $8 \times 8$  pixel pieces and the DCT taken after by Huffman or number juggling coding, is used in the ISO JPEG (joint photographic expert group) draft worldwide standard for photograph compression. The deficiency of this technique is the blocking (or tiling) made by a human being become visible rise high the compression percentage[14].

## F. Lossless Compression Methods

- **Run length coding:** Run length coding replaces information by a (length, value) pair, where "value" is the repeating and "length" is the number of iterations. This technique is particularly winning in pressing bi-level images since the incident of a long running of value is irregular in regular dim scale images. Intent to this is to decompose the dim scale image into bit planes and press singular bit-plane independently[14].

## G. Lossless Predictive Coding

Lossless predictive coding predicts the estimation of each pixel by using the estimations of its adjacent pixels. Thus, every pixel is encoded with an expected mistake rather than its true value. Generally, the mistakes are less important in contrast with the real value in order to use less bits are compulsory to store them.

- **DPCM (differential pulse code regulation)** is a prescient coding based lossless photo compression plot. It is excessively the base for lossless JPEG pressure. A variety of the lossless prescient coding is the versatile forecast that partitions the photographs into smaller sections and computes the expectation coefficients independently for each piece to achieve high forecast execution. It can to be merged with other plan to get a cross breed coding calculation with predominant execution.
- **Multi-determination Coding:** HINT(Hierarchical Interpolation) is a multi-determination coding technique in view of sub-samplings. It starts with a low-resolution publication of the real image, and inserts the pixel value

to successively create prevalent resolutions. The mistakes among the inserted values and the true values are put, along the 1<sup>st</sup> low-determination picture. Compression is achieved since both the low-determination image and the mistake values can be put along with lesser bits than the real image. Normally, the picture is reversibly adjusted into an arrangement of unique determination sub-images in multi-resolution coding. Usually, it diminishes the entropy of the picture. A few sorts of tree portrayal could be worn to get more compression by misusing the tree arrangement of the multi-resolution techniques[14].

**Merits and Demerits of Various Techniques**

Techniques	Merits	Demerits
JPEG[12]	This format is extremely portable. Coefficients are nearly uncorrelated.	It loses certain actual contents of the image. Impossible to completely decorrelate the blocks at their boundaries.
Wavelet Transform [12]	Higher Compression Ratio.	Coefficient quantization Bit allocation.
Vector Quantization Compression [13]	No coefficient Quantization.	Not appropriate for a low bit rate compression.  Slow codebook generation.
Fractal Compression [13]	Resolution-independent decoding property.	Slow Encoding.
Quantization [15]	Treating inputs as a single unit increases the optimality of the quantization.	It increases the cost and complexity of computation.
Transform Coding[14]	Many of the resulting coefficients for most natural images have small magnitudes and can be quantized without causing significant distortion in the decoded image. Reduce inter-pixel redundancies in the input image.	Lower quality of the original input.  Result may not be identical to the original input. Must hold all data in memory at once to perform transform.

Block Transform Coding[14]	All operations are independent; they can be performed in parallel to improve computational efficiency.	Blocking (or tiling) artifacts become visible at soaring compression ratios.
Run Length Coding[16]	Does not require much CPU horsepower. Easily implemented.	It is only efficient for lots of repetitive data.
Lossless Predictive Coding[17]	This coding uses the encryption of data so the data is secured until the destination.	Data gets faded if transmitted to the long distance.
Differential Pulse Code Modulation [14]	Good compression ratios. It reduces the bit rate for representing sample value.  Less bandwidth.	It is a complex process.

**V. Conclusions**

This review has demonstrated that the mammography approaches by including different picture compression strategies as depicted in this paper. Compression of picture can be enhanced by considering frequency domain redundancy. The effectiveness of frequency domain was observed to be exceptionally relying on individual picture sequences. Given the outcomes from prior work that discovered short-term prediction of the disease in the mammography to be more helpful for picture, however here we can presume that the generally poor performance of short duration, for a few sequences, is because of unearthly forecast being more productive than temporal.

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