

Performance of segmentation in infrared breast thermograms using level set method

S. Saran Raj ^{1*}, Hariharan. R ²

¹ Assistant Professor,

Department of Computer Science and Engineering, School of Computing, Vel Tech Rangarajan Dr. Sagunthala R&D Institute of Science and Technology, Avadi, Chennai-62, TamilNadu, India.

² Assistant Professor,

Department of Information Technology Vel Tech Rangarajan Dr. Sagunthala R&D Institute of Science and Technology, Avadi, Chennai-62, TamilNadu, India.

*Email: hharanbtech@gmail.com

Abstract

Breast cancer is the overall most basic obtrusive tumour in females. The breast cancer can be dealt with successfully that they are analyzed at a beginning time. It can watch the principal indication of shaping up disease before mammography can recognize. The thermal data can be appeared in a pseudo shading where each shading speaks to a particular scope of temperature. Different techniques can be connected to extricate hot districts for distinguishing associated areas with interests in the thermograms and possibly suspicious tissues. In this paper at first the pre process of the thermogram pictures are done then they are improved. The upgraded pictures are divided by two picture segmentation strategy: K-means and level set technique are study and compared. The highlights have been extracted and classification for both the segmentation techniques.

Keywords: Breast Thermograohy, Level Set, K-Means

1. Introduction

Cancer (tumor) is a class of diseases identify by uncontrolled proliferation of cells. Cancer cells have the capability to spread, either by invasion i.e., direct development into neighbouring tissue or by implantation into removed destinations by metastasis. [1] There are different types of cancer such as breast cancer, oral cancer, lung cancer, pancreatic cancer, bladder cancer, leukemia (blood cancer) etc. Women with regular breast screening have a considerable reduction in breast cancer mortality compared with women who do not. Cancer cells are classified as benign or malignant masses [2]. Human body whose temperature is above zero Kelvin has the ability to emit infra-red radiations as an exponential function of temperature. ideal body is an object which notice all radiations occurrence on it and release them back. The wavelength of emission is dependent on temperature [3] and spectral emissivity, e (1"). For ideal body $e=1$, grey objects $e < 1$. The spectral emissivity of any objects will be in the range of $0 < e < 1$. According to radiation law Planks, the radiation power and its wavelength distribution is given as,

$$w(\lambda, T) = \frac{2\pi hc^2}{\lambda^4} \left[\exp \frac{hc}{\lambda kt} - 1 \right]^{-1} W cm^{-2} \mu m^{-1} \quad (1)$$

where

h (Planck's constant) = 6.6256×10^{-34} J s

c (velocity of light) = 2.9979×10^8 ms⁻¹

k (Boltzmann's constant) = 1.38054×10^{-23} W s K⁻¹

λ (Wavelength) in μm

T (Temperature) in K.



Fig 1. Infrared Thermography camera

Thermograph's measurement of heat emission from the tissues was investigated in the 1970s and 1980s, it was found to be of low specificity and selectivity. Heat transfer may be due to conduction in the tissues layers, convection by vascular system and radiation because of environment [4]. The heat patterns indicate the metabolic activity of breast, sweating and blood circulation in breast tissue. Temperature will be high at the region of the cancerous tissue compared to normal tissues in breast. However, as the picture only shows the surface temperature, clinician find problems in specifying the location, shape and type of the cancer cell which directly affect the decision for further therapy.

In this paper the proposed framework is discussed which contain the pre-processing method, segmentation methods are re-viewed, feature extraction and classifying is performed [5]. In the end the final output and conclusion are given to evaluate the comparison of the two segmentation methods.

2. Methodology

The work flow as follows: The thermography picture of breast is taken as an input. The quality of the picture we use enhancement techniques to improve which is adaptive histogram equalization. The Enhanced picture is then used as an input for segmentation.

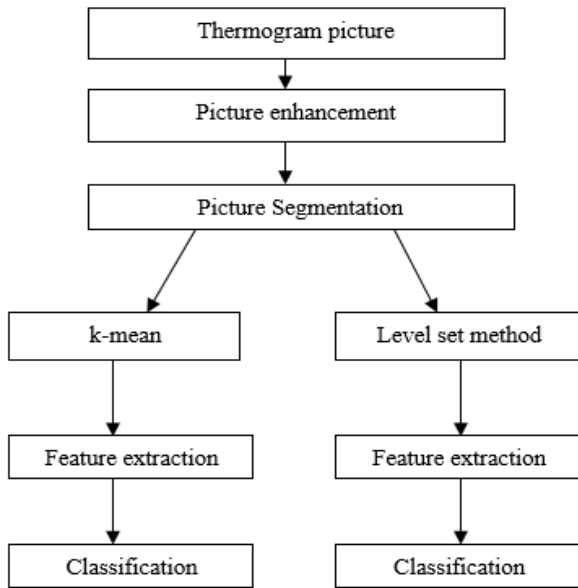


Fig 2. Methodology

2.1. K-means clustering

Kmean colour mapping segmentation is used for identifying the high temperature region on the breast. The main aim is to set k centroids, each contains one cluster. These centroids endure be in a foxy way since the various location gives various result, hence it is good to keep them very distance from each other [6].

The first set up is to take all point belongs to a given data base and relate it to the neighbour centroid value. At that point we have to re-calculate all k new centroids. Then result, k centroids may changes its location one by one in each cycle until no further modification are done.

$$j = \sum_{j=1}^k \sum_{i=1}^x ||x_i^j - c_j||^2 \tag{2}$$

where $||x_i^j - c_j||^2$ is a distance between a data point $x_i^{(j)}$ and the cluster centre c_j , is an distance of the n data points from their particular cluster centres.

Then the k-mean algorithm is used to perform the following steps below the convergence. Iterate until it attains the stable condition:

1. The centroid coordinate to be determined.
2. Distance of each object to the centroid.
3. Group the minimum distance object.

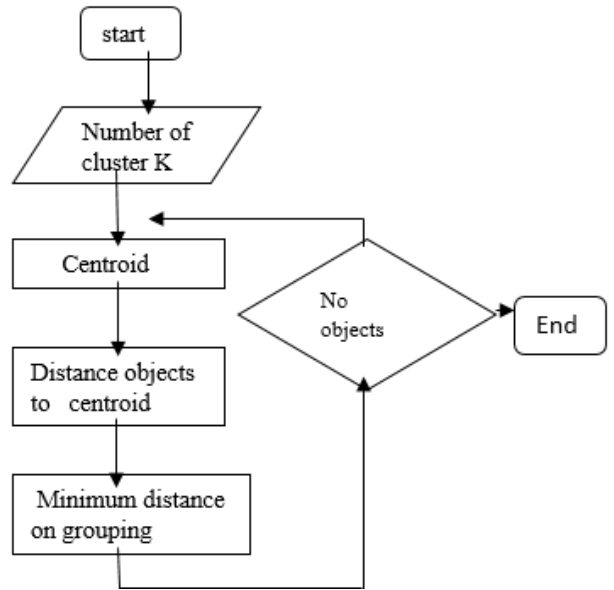


Fig 3. K-Mean Algorithm

2.2. Level set method

Active contour is an deformable spline determined by, minimizing of energy constraint and picture forces. They're usually utilized in picture search and pc vision to discover object margins, and to explain their form. The foremost idea is to interchange the curve at intervals in an picture with the speed F towards options of interest like edges and contours. However, the method become extremely complex and difficult if the active contours attempt to cross over themselves, or if its form attempts to interrupt into two.

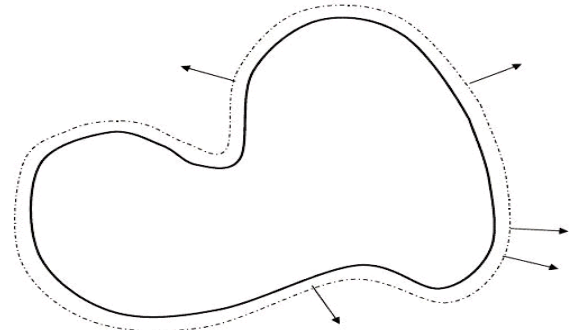


Fig 4. Active contour evolves under influence of forces

The problems related to energy reduction are overcome by introducing a new structure for active contour structure. In this methodology the location of contour does not change but only the points values changes. As a concern, this methodology gives a substantial feature for level set methodology which tends to handle topological combining and dividing. [7]

The partial differential equation can be described by satisfying the level set function ϕ performed in the standard path along with speed function F.

$$\frac{\partial \phi}{\partial T} + F|\nabla \phi| = 0 \tag{3}$$

This Hamilton- Jacobi equation is said to be known as level set evolution equation and that guides the motion of the LSF. Segmentation with Level Set consists of:

1. To Initialize the level set function on the initial contour C_0 . This function can be anything as long as its zero level set arrives its contour value. [8]

$$c_0(x, y) = \{(x, y) | \phi(x, y, 0) = 0\} \tag{4}$$

2. Then the updated function in the narrow band is described based on Hamilton-Jacobi equation. [9].
3. This simulated iteration over step 2 until predefined number of iterations are reached or until the required object has been extracted. [10].

LSF is considered as a signed distance function (SDF) which [11] is said to be negative for within and acts as positive outside the contour. However, this amount set evolution of equation doesn't guarantee to protect the LSF as a signed distance perform that cause instability and numerical errors in the execution of the process. For picture segmentation, the speed term formulation is predicated on two dissimilar basis and it is classified into two methods the first as main classes exist for level set method and the next is edge-based models and region-based models. Edge-primarily based[12] models propagate the contour supported by the information provided on the basis on picture gradients and object boundaries, and also the rate of speed to attain zero price because it gets nearer to the object boundaries. Textural characteristics are extracted by picture processing techniques is based on the Gray Level Co occurrence Matrix (GLCM) which is extracted from each picture. [13] Co-occurrence matrices can also be calculated for four different directions are 0°, 45°, 90° and 135° degrees. These specific characteristic is specifically used to train the SVM classifier. Then the developed classifier is tested for different thermogram Pictures.

3. Results

The suggested method is implemented by using Matlab software and the breast thermogram picture datasets are given as the input. The output are shown in figure.

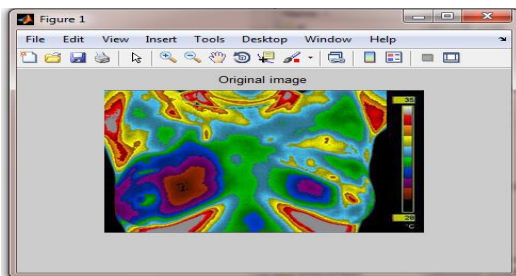


Fig 4. Input thermogram picture

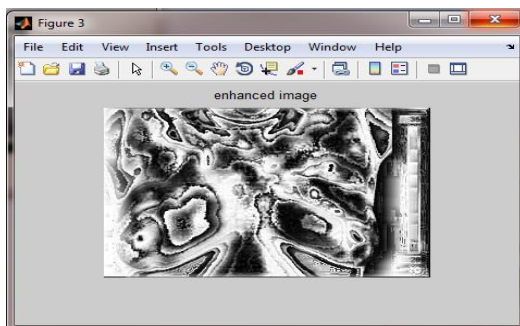


Fig 5. Enhanced picture after grayscale conversion

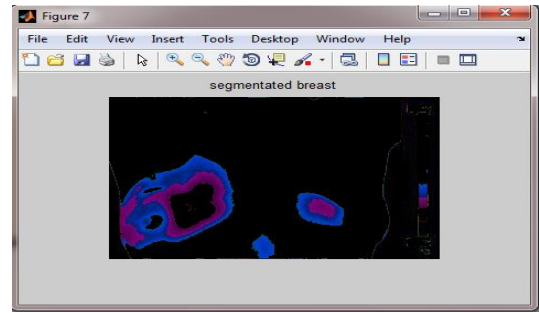


Fig 6. K means segmentation using colormap result

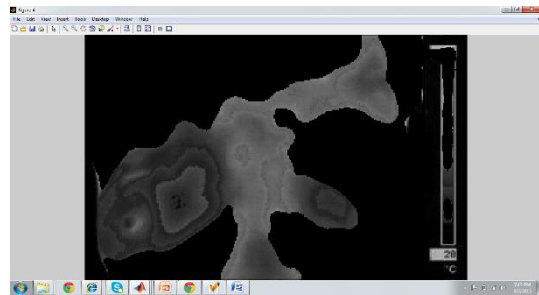


Fig 7. Level set segmented result

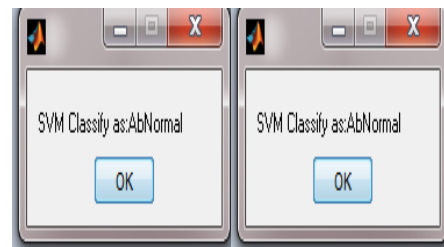


Fig 8. Classifier outputs

Table 1: parameter comparison for SVM

Parameter(%)	K-Mean	Level Set
Accuracy	89.71	91.63
Sensitivity	92.21	93.60
Precision	92.21	92.89

4. Conclusion

Using level set methodology and k-means, the paper clearly aims on the division of carcinoma from thermogram pictures. Quality of picture is increased with help of adaptive histogram deed and so they're divided. Thermogram pictures are not only using for breast and also to find the defect in others parts of body. The methodology used should differentiate Based on the applications.

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