

# SVM Classifier Based Automated whole Breast Ultrasound Screening

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**Abstract**— This paper provides the derivation of speckle reducing anisotropic diffusion (SRAD), a diffusion method tailored to ultrasonic and radar imaging applications. SRAD is the edge-sensitive diffusion for speckled images, in the same way that conventional anisotropic diffusion is the edge-sensitive diffusion for images corrupted with a additive noise. The proposed method has been anisotropic diffusion Speckle detection. In resolution quality by means of using anisotropic diffusion Speckle detection. In the presence of speckle noise, speckle reducing anisotropic diffusion excels over the traditional speckle removal filters and over the conventional anisotropic diffusion method in terms of means preservation, variance reduction, and edge localization.

**Keywords**— Anisotropic diffusion, image enhancement, speckle reduction, ultrasound imaging, Svm, watershed segmentation.

## 1. INTRODUCTION

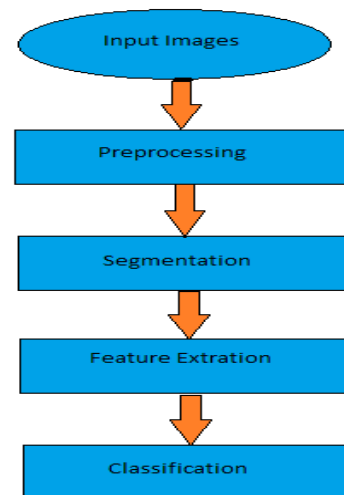
**SPECKLE**, a form of multiplicative, locally correlated noise, plagues imaging applications such as medical ultrasound image interpretation. For images that contain speckle, a goal of enhancement is to remove the speckle without destroying important image features. In certain applications, however, the removal of speckle may be counterproductive. We using watershed algorithm describes the specifics of the implementation and segmentation. We design the new multi-coordinate HOG (MCHOG) descriptors to accommodate the possible rotation. Support vector machine using to get normal or abnormal images.

## 2. ANISOTROPIC DIFFUSION VS ADAPTIVE SPECKLE FILTER.

### A. Anisotropic Diffusion

The filter also strikes a balance between averaging and the all-pass filter. In this case, the balance is achieved by forming an exponentially shaped filter kernel that can vary from a basic average filter to an identity filter on a point wise, adaptive basis. Again, the response of the filter varies locally with the coefficient of variation. The Provide Frost filter also strikes a balance between averaging and the all-pass filter. In this case, the balance is achieved by forming an exponentially shaped filter kernel that can vary from a basic average filter. otherwise coefficient of variation, the filter is more average-like, and in cases of high coefficient of variation, the filter attempts to preserve sharp features by not averaging. When the coefficient of variation exists in between the two thresholds, a balance between averaging and the identity is computed (as with the standard Lee and Frost filters).

### B. system design.



## 3. WATERSHE ADLGORITHM.

Over all watershed algorithm derives its name from the manner in which regions are segmented into *catchment basins*. The choice of height function depends on the application; the basic algorithm is independent of the height function. Therefore, the steps of the watershed segmentation algorithm are as follows:

1. Compute the curvature (or some other height function) at each vertex.
2. Find the local minima and assign each a unique label.
3. Find each flat area and classify it as a minimum or a plateau.
4. Loop through plateaus and allow each one to descend until a labeled region is encountered.

5. Allow all remaining unlabeled vertices to similarly descend and join to labeled regions.

6. Merge regions whose watershed depth is below a preset threshold for further concept.

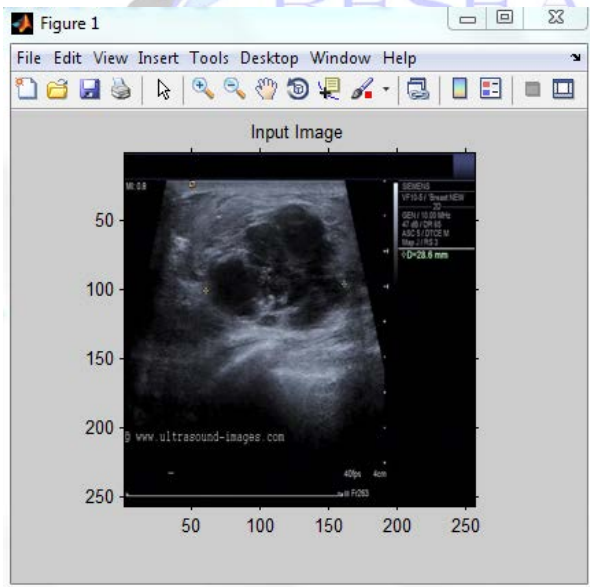
*A. Oversegmentation.*

This section demonstrates the need for region merging by displaying some results obtained without performing the final merging step. The output of the watershed algorithm suffers quite badly from over segmentation's .Any small defects or noise present in the initial model allow the formation of bumps on the surface. If these bumps are concave, they in turn lead to the creation of local minima, which become isolated regions as neighboring vertices flow into them.

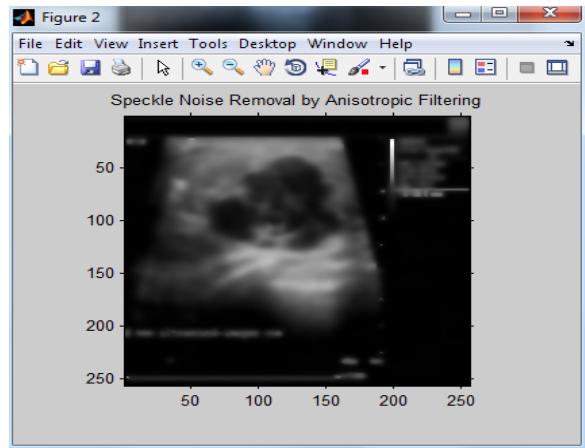
**4. FIGURES AND TABLES.**

Using speckle Reducing Anisotropic Diffusion filter for get noisiest images (preprocessing).The Frost filter also strikes a balance between averaging and the all-pass filter. In this case, the balance is achieved by forming. an exponentially shaped filter kernel that can vary from a basic average filter to an identity filter on a point wise, adaptive basis.

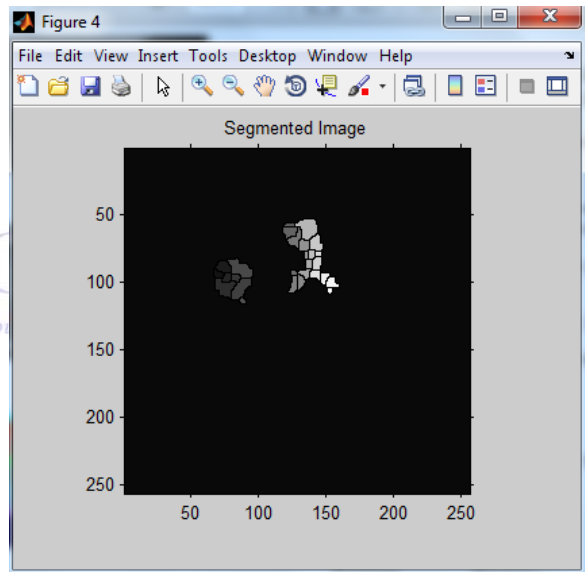
*A. Input Image.*



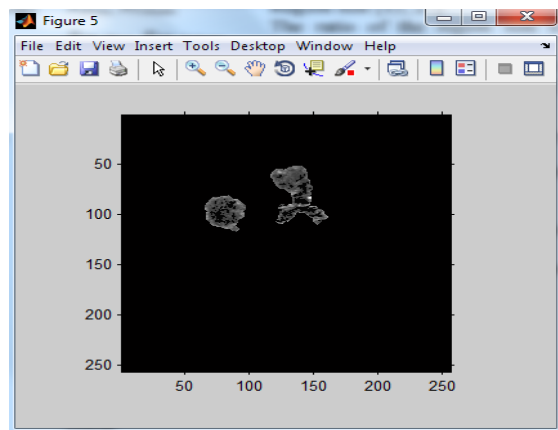
*B. Noise Removal.*



*C. Segmented Image.*



*D. Extracted Abromality.*



E. precisions recall graph.

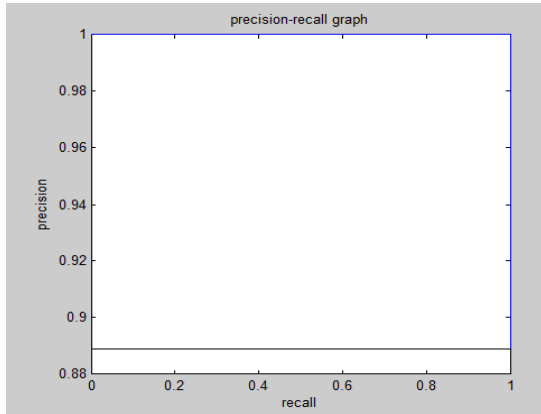


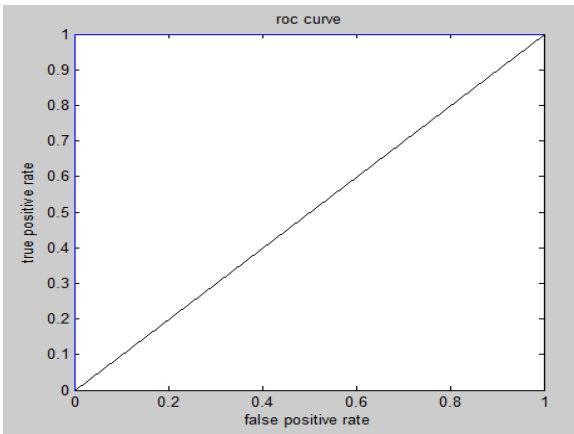
TABLE I.

Sensitivity (%)	Malignant (cm)				Total	Benign (cm)				Total
	<1.0	1.0-2.0	2.0-3.0	>3.0		<1.0	1.0-2.0	2.0-3.0	>3.0	
50	4	12	2	11	29	10	18	5	4	37
60	2	11	0	8	21	9	15	4	4	32
70	1	10	0	8	19	6	11	1	3	21
80	1	7	0	4	12	4	6	1	3	14
90	0	3	0	3	6	2	1	1	3	7

TABLE II

Fig. 1. IMPROVE SENSITIVITY OF THE SYSTEM

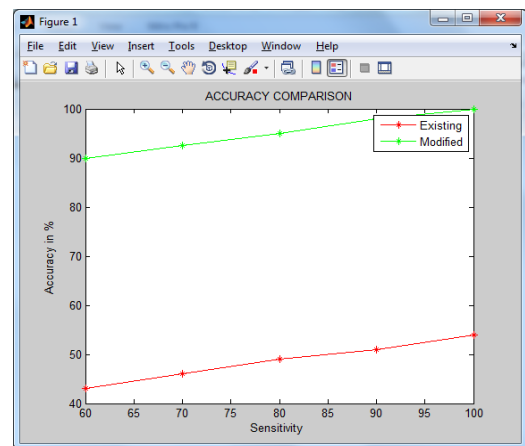
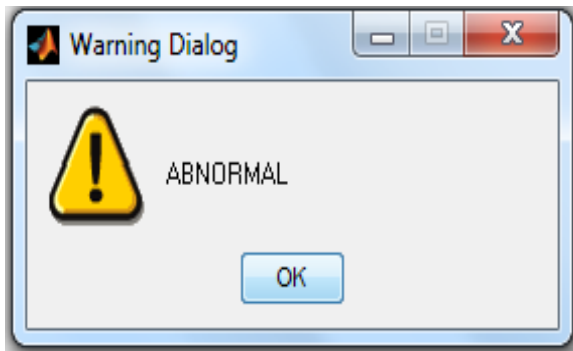
F. ROC Curve.



SENSITIVITY IN %	EXISTING	PROPOSED
60	43	90
70	46	92.5
80	49	95
90	51	98
100	54	100

Fig 2. CLASSIFICATION ACCURACY COMPARISON GRAPH.

G. classifier result.



## 5. CONCLUSION.

The motivation of this study focused on designing a system that would allow us to find the tumor detection. The system potential detect tumor Images efficiently and effectively by using Support Vector Machine (svm), anisotropic diffusion, watershed segmentation Algorithm and mult coordinate hog (HOG). Based on the target, the sensitivity of 100% should be the performance index for the detection ability. Detecting and providing all tumors in the ABUS images for radiologists to do malignancy evaluation would be better than only the potential malignant tumors suggested by CAD systems .Our model had to test for the presence of a potential detecting and providing all tumor in the ABUS images for radiologists to evaluation would be better An ABUS image explorer and marking system is going to be constructed for radiologists to mark tumors and review the abnormalities suggested by the proposed CAD algorithm. With the quantitative system, recording tumor locations and generating examination report for ABUS database would be more efficient and sensitive effectively.

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