

Home Page

Print

Title Page

Contents



Page 1 of 13

Go Back

Full Screen

Close

Quit

MAMMOGRAM MASS CLASSIFICATION BASED ON DISCRETE WAVELET TRANSFORM TEXTURAL FEATURES

Authors:

J. Abdul Jaleel, Sibi Salim, Archana. S
Dept. of Electrical & Electronics Engineering
TKM College of Engineering, Kollam
Kerala

25 /09/2014

Home Page

Print

Title Page

Contents



Page 2 of 13

Go Back

Full Screen

Close

Quit

Contents

1	OBJECTIVE	3
2	METHODOLOGY	4
3	IMAGE DATABASE	5
4	SEGMENTATION	6
4.1	Manual Segmentation	6
5	FEATURE EXTRACTION	7
5.1	Discrete Wavelet Transform	7
6	CLASSIFIERS	8
6.1	K Nearest Neighbor classifier	8
6.2	Support Vector Machine (SVM)	9
6.3	Radial Basis Function Neural Network(RBFNN)	10
7	Performance Comparison	11
8	CONCLUSION	11
9	REFERENCES	12

Home Page

Print

Title Page

Contents



Page 3 of 13

Go Back

Full Screen

Close

Quit

1. OBJECTIVE

- To develop a Computer Aided Diagnostic (CAD) system based on discrete wavelet transform textural features using K- Nearest Neighbor classifiers (K- NN), Radial Basis Function Neural Network (RBFNN) & Support Vector Machine (SVM) classifying breast nodules as either benign or malignant on mammographic images.

2. METHODOLOGY

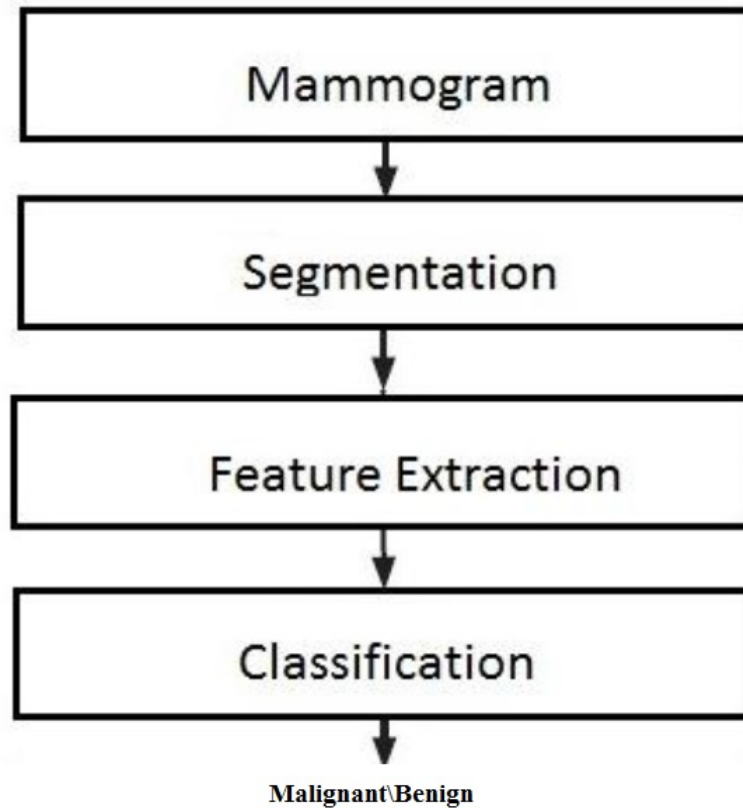


Figure 1: Methodology of the developed CAD system

3. IMAGE DATABASE

- Mammogram is a low dose x- ray exam of the breasts to look for any abnormalities.
- The results are recorded on x- ray film (Screen Film Mammograms) or directly into a computer (Digital Mammograms).
- Images used are Digitized Screening Mammograms.
- Mammograms in the Mini Mammographic Image Analysis Society (MIAS) database is used here.
- The size of all the images is 1024 pixels x 1024 pixels.
- Image Database with 148 images: 56 Malignant and 92 Benign images.

4. SEGMENTATION

4.1. Manual Segmentation

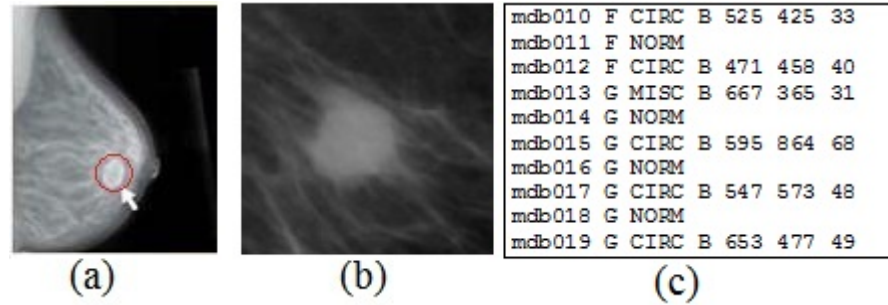


Figure 2: (a) Mammogram with circular ROI. (b) Manually segmented ROI. (c) Database Details

- Mini MIAS database provides the ground truth information with each mammogram.
- Suspicious region is specified in terms of x, y co ordinates with r radius by the radiologist: circular ROI.
- Region of Interest (ROI) is manually segmented as rectangular region of interest.

$$I_{AD} = [x - r, 1000 - y] \quad (1)$$

$$I_{NonAD} = [x + 50 - r, 900 - y] \quad (2)$$

5. FEATURE EXTRACTION

5.1. Discrete Wavelet Transform

- Energy measures textural uniformity.

$$Energy = \sum_{x,y} (d_{x,y}^{subband})^2 \quad (3)$$

- Kurtosis is a measure of whether the data are peaked or flat relative to a normal distribution.

$$\beta = \frac{E[(g - \bar{g})^4]}{(E[(g - \bar{g})^2])^2} \quad (4)$$

- Standard deviation is a measure of contrast in an image.

$$\sigma = \sqrt{\sum_{g=0}^{L-1} (g - \bar{g})^2 P(g)} \quad (5)$$

6. CLASSIFIERS

6.1. K Nearest Neighbor classifier

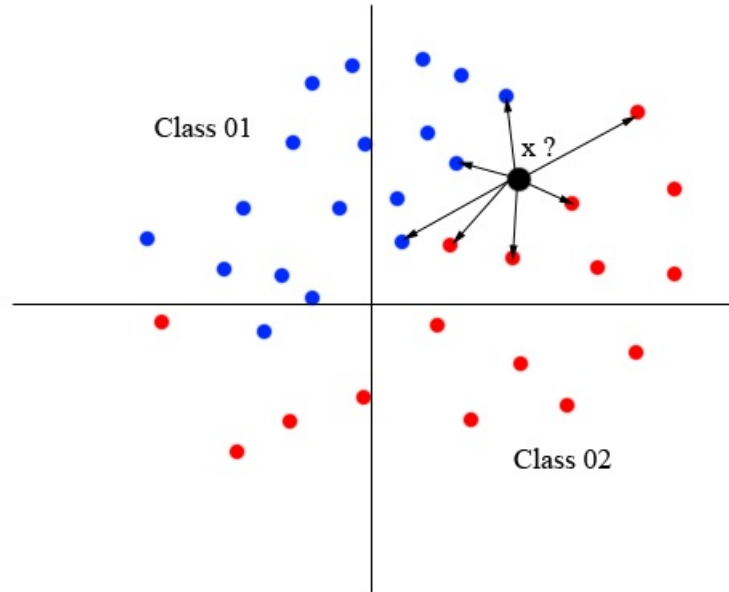


Figure 3: k-Nearest Neighbor algorithm representation

- Distance can be calculated using:

$$d(x_i, x_j) = \sqrt{\sum_{r=1}^n (x_{ir} - x_{jr})^2} \quad (6)$$

Home Page

Print

Title Page

Contents

◀◀

▶▶

◀

▶

Page 8 of 13

Go Back

Full Screen

Close

Quit

6.2. Support Vector Machine (SVM)

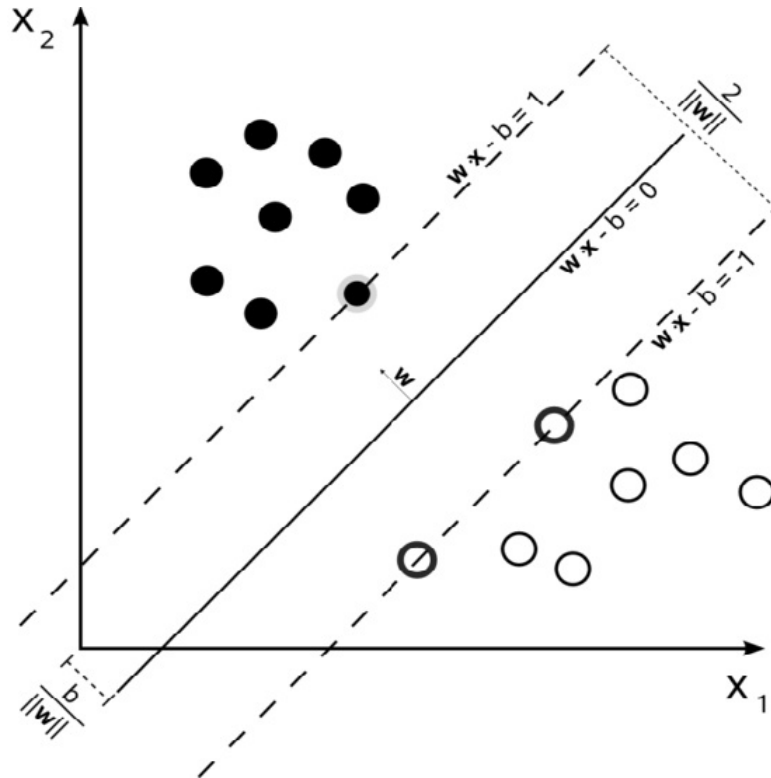


Figure 4: SVM classification general representation

6.3. Radial Basis Function Neural Network(RBFNN)

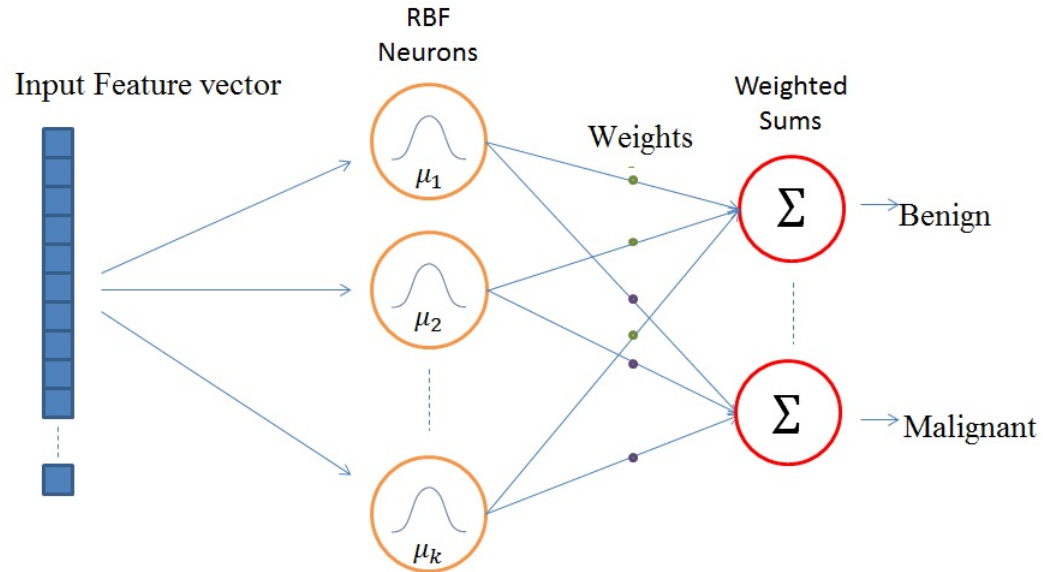


Figure 5: RBFNN architecture

- Input feature vector, Hidden layer with 32 RBF neurons & output layer with 2 neurons.

7. Performance Comparison

K-NN	P'	N'	SVM	P'	N'	RBFNN	P'	N'
P	37	9	P	42	4	P	43	3
N	1	27	N	3	25	N	5	23

Figure 6: Performance Analysis

8. CONCLUSION

- Segmentation of mammograms were done with manual segmentation and feature extraction is done by DWT.
- The accuracy of K-NN classifier can be varied from 80% to 85% by changing the value for k.
- SVM Classifier has given an average accuracy of 85% for DWT features.
- RBFNN with DWT features outperforms all other classifiers with an accuracy of 93.24%.

9. REFERENCES

- [1] Rangaraj M. Rangayyana, Fabio J. Ayresa, J.E. Leo Desautels, "A review of computer-aided diagnosis of breast cancer: Toward the detection of subtle signs", Journal of the Franklin Institute 344 (2007).
- [2]I. Christoyianni, A. Koutras, E. Dermatas, G. Kokkinakis, "Computer aided diagnosis of breast cancer in digitized mammograms", Computerized Medical Imaging and Graphics 26 (2002)
- [3]K. K. Rajkumar, G. Raju, "Classification of mammogram images using discrete wavelet transformations", Advances in computing and communications in computer and information science. vol.192, 2011, pp.435- 443.
- [4]M. Kociolek, A. Materka, M. Strzelecki, P. Szezypinski, " Discrete wavelet transform derived features for digital image texture analysis", Proc. of international conference on signals and electronic systems, 2001, pp. 163-168.
- [5]M. A. Alolfe, A. M. Youssef, Y. M. Kadah, A. S. Mohamed. "Computer aided diagnostic system based on wavelet analysis for microcalcification detection in digital mammograms", IEEE Proc.CIBEC, 2008.

Home Page

Print

Title Page

Contents

◀

▶

◀

▶

Page 12 of 13

Go Back

Full Screen

Close

Quit

Home Page

Print

Title Page

Contents



Page **13** of **13**

Go Back

Full Screen

Close

Quit

THANK YOU