Symmetrical Cluster Analysis for Thermographic Breast Cancer Detection

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OBJECTIVE

This study aims to develop an algorithm that can effectively and accurately distinguish between tumor-affected and normal breast tissue. This is being achieved by performing statistical and cluster analyses on the existing dataset to identify key characteristics of tumor-affected tissues.

INTRODUCTION

Thermographic Imaging utilizes an infrared thermal camera to capture the skin temperature, and potentially indicate tumorous regions of interest based on the skin temperature patterns. Prior research indicates that tumorous regions are warmer than normal breast tissue, and that tumorous tissue cools at a slower rate than normal breast tissue [1].

MATERIALS & METHODS

- Patients and Volunteers were imaged over the course of 15 Minutes with an N2 Infrared Camera with Thermal Resolution of 50 mK / Digital Count.
- Image Processing and Algorithm Development were completed in MATLAB.

Image Analysis was carried out with two mechanisms:
(1) Statistical Analysis of tumorous regions as indicated by truth data.
(2) Cluster Isolation and Symmetric Removal to identify potential tumorous regions.

Statistical Analysis was carried out on both Tumorous and Normal tissue by comparing the Region of Interest and the Opposing Region. Each region was identified using the truth data given and user manual selection of patient nipples as reference points. The pixel intensity levels of each region in the first and last image of the series were statistically analyzed. Based on this data, a pixel intensity threshold was set and spatial clusters were formed using the DBSCAN Clustering Method [2]. Clusters were then isolated using a symmetrical analysis algorithm to identify clusters of clinical value.

RESULTS

ROI and OPP Region Analysis results displayed are for Patient IRST011, one of 14 patients imaged.

Figure 1: Patient First Image with Region of Interest (Red) and Opposite Region (Green) highlighted.

Figure 2: Patient Final Image with Region of Interest (Red) and Opposite Region (Green) highlighted.
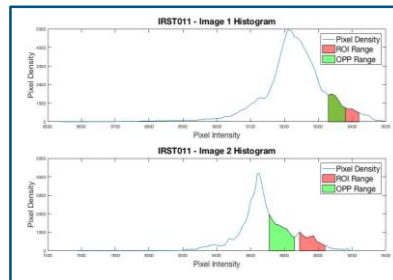

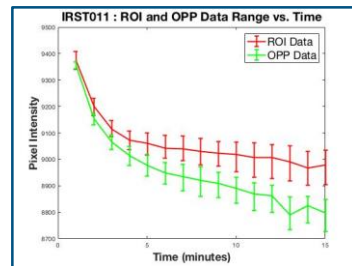**Figure 3:** Pixel Density graph with ROI range and OPP range highlighted

Figure 4: Mean Intensity of ROI and OPP throughout imaging window. Intensity Range for ROI and OPP shown at each data point.

RESULTS (Cont.)

The Cluster Isolation Algorithm was tested on both volunteers and patients. The volunteers were assumed to have no tumorous regions and the patients have
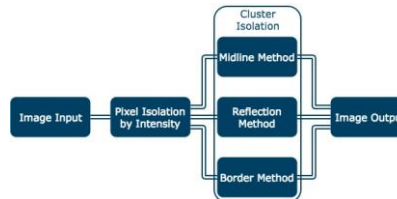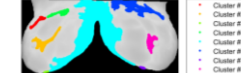

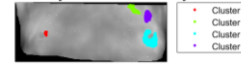**Figure 5:** Block Diagram for Cluster Isolation Algorithm
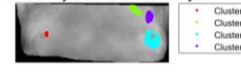
Figures 6-9: Image Inputs and Outputs to Cluster Analysis Algorithm for Patients 8 and 11
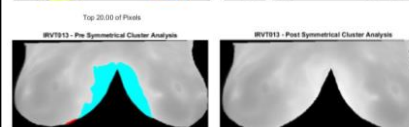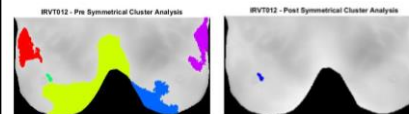
Figures 10-14: Cluster Analysis Algorithm Performed on Volunteers 12 and 13

DISCUSSION & FUTURE WORK

Results from the ROI and OPP Region Analysis confirm findings from prior research that tumor-affected breast tissue is warmer than normal tissue.

For each of the patients evaluated, the tumor region identified by the truth data was significantly warmer than the corresponding region on the opposite (unaffected) breast with a 95% confidence level. The tumorous region remained warmer than the normal tissue over the entire period of imaging.

Given these results, the cluster isolation algorithm becomes more significant due to its ability to isolate unilateral regions that are warmer than the surrounding tissue.

As more patients are added to the data set, we will continue to train the algorithm. The goal is to provide information for adjunct usage with mammography that may improve the overall accuracy of early breast cancer diagnosis.

CONCLUSION

Given the available data from the 14 patients images thus far, our research indicates that the tumorous breast tissue is noticeably warmer than normal breast tissue. This is evident through the comparison of the tumor-affected tissue with the corresponding region on the opposite breast, which acts as patient-specific baseline. Additionally, these results allow us to train the Cluster Isolation algorithm to identify regions of clinical importance.

REFERENCES

[1] Li Jiang, et al., Phys. Med. Biol. 56 (2011). 187–202
[2] Yarpiz (2015). DBSCAN Clustering Algorithm, MATLAB Central File Exchange. Retrieved July 2017.