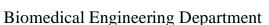


### بسم الله الرحمن الرحيم

# Sudan University of Science and Technology College of Engineering





A dissertation submitted in partial fulfillments of the requirements for the B.Sc. Degree in Biomedical Engineering

# **Breast Cancer Detection in Mammographic Image**

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**August -2014** 

### **Abstract**

Breast cancer is one of the major causes of death in women when compared to all other cancers. This research was developed methods that can be used to detect abnormal breast tissues. Matlab programs were devolved to detect abnormal tissue in mammograms, to assist radiologist in their diagnosis as a second opinion. In this study100 image consist 100 normal ROIs, 100 abnormal ROIs (contained 61 benign ROIs and 39 ROIs malignant ROIs) was extracted from digital mammogram which obtained from mini-MAIS database. Five First order statistical features were calculated for each ROI. Fourteen Haralick features according to GLCM matrix was calculate for each ROI. The best features were selected by their ability to distinguish between normal and abnormal tissue to use it in program to detect breast cancer.

# المستخلص

سرطان الثدي واحد من أكبر أسباب وفاة النساء في العالم مقارنة مع باقي السرطانات. هذا البحث طور طرق تستخدم لعمل كشف لخلايا الثدي المريضة. برامج ماتلاب تم تطوير ها لعمل كشف الخلايا الغير طبيعية في صور الماوقرام الرقمية لمساعدة أخصائي الأشعة في تشخيصه كخيار ثاني. في هذه الدراسة تم استخدام 100 صورة تحتوي على 100 عينة سليمة و100 عينة مريضة (تحتوي 61 عينة حميدة و39 عينة خبيثة) AMIAS استخراجها من صور الماموقرام الرقمية التي تحصل عليها من جمعية تحليل صور الماموقرام ( ) Haralick خمس معادلات من الدرجة الأولى تم حسابها لكل العينات .أربعة عشر من معادلات هارليك ( ) تم حسابها لكل العينات. أفضل المعادلات تم اختيار ها على GLCMالتي تعتمد على المصفوفة الأحصائية ( ) أساس قدرتها على التمييز بين الخلايا السليمة والمريضة .وأستخدمت في برنامج للكشف عن سرطان الثدي.

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# **CHAPTER ONE**

Introduction

# 1.1 General View:

Breast cancer is one of the major causes of death in women when compared to all other cancers. Studies have demonstrated that African women are more prone to get breast cancer at an early age as compared to Western women. One study showed a high percentage occurred in Sudanese women less than 50 years of age showing a clear indication of early onset of the disease in younger women. This is especially true in the case of Sudanese women. Early detection is the key to control the breast cancer and decrease the mortality rate. Mammography is one of the imaging modality in early breast cancer detection typically through detection of characteristic masses and micro calcifications. There is variation and inconsistent between radiologists in diagnose the breast cancer. The radiologist may lose part of the tumor or faulting in determining the location, number and size of the tumor [1].

Computer-aided detection (CAD) tool developed to aid radiologist in detectingtumors. CAD system could act as a second opinion, it also offer better accuracy assists in reducing missed cancers and provide better prognosis [2]. There are many technique that used to developing CAD system. One of these techniques is features extraction from Regions of Interest (ROI) of Mammograms. Several types of features extraction from digital mammograms including position feature, shape feature and Texture feature. Textures are one of the important features used for description, analysis and classification of ROI in Mammograms. The texture features are ability to distinguish between abnormal and normal cases. Texture measures are two types, first order and second order. In the first order, texture measure are statistics calculated from an individual pixel and do not consider pixel neighbor relationships. Intensity feature are first order texture calculation [3]. In the second order, measures consider the relationship between neighbor pixels GLCM is a second order texture calculation [4]. Texture features has been extracted and used the best features that can help to detection abnormal breast cancer and distinguish between normal and abnormal breast tissues.

# **1.2 Problem Statement:**

There is variation and inconsistent between radiologists in diagnose the breast cancer due to factoversvision, experience and skill of the Radiologist, the image quality and the highamount of cases that are being tested and diagnosed during the day. The radiologist may lose part of the tumor or faulting in determining the location, number and size of the tumor.

### 1.3 Objectives:

## 1.3.1 General aim:

Developing CAD system that used as a second opinion of diagnosing Breast Cancer.

### 1.3.2 Specific objectives:

- 1. Collect mammogram Data.
- 2. Design code to read image and select ROI.
- 3. 1<sup>st</sup> order Feature extraction from ROIs.
- 4. Gray Level Co-occurrence matrix (GLCM) calculate and Normalization of GLCM.
- 5. 2<sup>nd</sup> order feature extraction (Haralick).
- 6. Choose the best features to distinguish between normal and abnormal tissues

# 1.4 Thesis Layout:

This project contains six chapters. Chapter one contains an introduction to the project, chapter two about literature review that related with the project, while chapter three contains a theoretical background for the project components, chapter four contains materials and methods used to achieve the objectives of the project, the results and discussion are found in chapter five, and finally chapter six contains conclusion and recommendation.