

Multimodality Breast Imaging Diagnosis and Treatment







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Preface

Breast cancer is an abnormal growth of cells in the breast, usually in the inner lining of the milk ducts or lobules. It is currently the most common type of cancer in women in developed and developing countries. The number of women affected by breast cancer is gradually increasing and remains as a significant health concern. Hence, the early detection of breast cancer can improve the survival rate and quality of life. Therefore, today, newer modalities are available to more accurately detect breast cancer. Researchers are continuously working to develop novel techniques to detect early stages of breast cancer. This book covers breast cancer detection using different imaging modalities such as mammography, magnetic resonance imaging, computed tomography, positron emission tomography, ultrasonography, infrared imaging, and other modalities.

Architectural distortion is one of the major causes of false-negative findings in the detection of early stages of breast cancer. Chapter 1 presents methods for computer-aided detection of architectural distortion in mammograms acquired prior to the diagnosis of breast cancer in the interval between scheduled screening sessions. The results are promising and indicate that the proposed methods can detect architectural distortion in prior mammograms taken 15 months (on average) before clinical diagnosis of breast cancer, with a sensitivity of 0.8 at 5.2 false positives per patient.

A computer-aided system for the automated detection of normal, benign, and cancerous breasts using texture features extracted from digitized mammograms and data mining techniques is proposed in Chapter 2. The novelty of this work is to automatically classify the mammogram into normal, benign, and malignant classes using the texture features alone, with an efficiency of 93.3% and sensitivity of 92.3% using a fuzzy classifier.

Breast cancer diagnosis by combination of fuzzy systems and an ant colony optimization algorithm is proposed in Chapter 3. Results on the breast cancer diagnosis dataset from the University of California Irvine machine learning repository show that the proposed FUZZY-ACO would be capable of classifying cancer instances with a high accuracy rate and adequate interpretability of extracted rules.

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Chapter 4 discusses a computer-aided diagnosis system tested on magnetic resonance datasets obtained from different scanners, with a variable temporal and spatial resolution and on both fat-sat and non–fat-sat images, and has shown promising results. This type of system could potentially be used for early diagnosis and staging of breast cancer to reduce reading time and to improve detection, especially of the smaller satellite nodules.

Imaging plays a pivotal role in the evaluation of metastatic spread of breast cancer disease. Chapter 5 gives an overview of the recent developments in breast cancer imaging, in terms of instrumentation and clinical applications. In addition, the theoretical framework behind advanced imaging modalities is highlighted to provide background knowledge to the reader, and potential future research directions are also presented.

The role of positron emission tomography is established in the practice of oncology. The advances in functional and molecular imaging techniques have increased the accuracy in the diagnostic evaluation of breast cancers and is discussed in detail in Chapter 6.

Chapter 7 discusses 3D whole-breast ultrasonography, which can provide the entire breast anatomy for later review. The 3D whole-breast ultrasound procedure and the training time are simpler and shorter than the traditional 2D US. It also provides interoperator consistency, and its reproducibility is better for follow-up studies.

Recent progress in medical ultrasound has paved the way for the evaluation of breast cancer. State-of-the-art high-resolution ultrasound can detect tiny breast lesions as small as 1–2 mm in size, and sometimes microcalcifications even less than 0.5 mm, or small carcinomas 3–6 mm in diameter. Chapter 8 presents an overview of the recent developments in ultrasound imaging of breast cancer, in terms of instrumentation and clinical applications.

Nonlinear features such as Lyapunov exponents are used to differentiate malignant and benign breast thermograms in Chapter 9. This work can be extended for classifying different stages of breast cancer. The authors are currently working toward these objectives.

A set of image features describing bilateral differences between left and right breast regions in thermograms is described in Chapter 10. These features are then used in a pattern classification stage to discriminate malignant cases from benign ones. Classification is performed by fuzzy if-then rules and applies a genetic algorithm to optimize the rule base, and secondly uses an ant colony optimization classification algorithm. Both approaches have shown good classification accuracy.

Infrared imaging has shown to be a promising technique for the early diagnosis of breast pathologies and as a screening technique. The concept of a combined diagnostic enables a high degree of specificity and sensibility in such diagnosis. Chapter 11 presents a concept of merging







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information from the images with other modalities of examination, such as mammograms and ultrasound, in order to improve the early detection of breast pathologies, including cancer.

Chapter 12 discusses diffuse optical imaging, which makes use of diffuse light to probe deep tissues by taking advantage of low tissue absorption within the near-infrared wavelength range (650–900 nm). The optical measurements obtained can be used to calculate optical properties, namely absorption and scattering within tissues. This, in turn, can provide information about physiological parameters within tissues, such as oxy- and deoxy-haemoglobin, and water and lipid, all of which can be utilized in the detection, characterization, and therapy monitoring of breast cancer.

Cytopathology is a branch of pathology that studies and diagnoses diseases on the cellular level, using samples of free cells or tissue fragments. Chapter 13 describes the results of a study of the features that are used by physicians and computers to diagnose cancer based on features in fine-needle aspiration cytology images. It discusses the significance of a cytological feature in representing its true ability to discriminate benign and malignant conditions of a breast lump in the Wisconsin Diagnostic Breast Cancer database.

Only a small number of studies have been reported on breast radiofrequency ablation, and most of them have included the posterior surgical excision of the treated breast. Chapter 14 presents the future trends in the development of more-specific radiofrequency algorithms for breast cancer treatment, to improve the results, determine the setting of the specific indications for the technique, and expand the study of long-term results and survival.

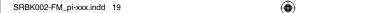
Breast conserving therapy is the gold-standard option for patients with early-stage breast cancer. The surgical excision removes the entire tumor with a negative surgical margin and helps to preserve the breast tissue as far as possible. Chapter 15 explains minimally invasive ablative techniques, which may offer complete tumor ablation, with less psychological morbidity, better cosmetic results, and shorter hospital stay.

A microwave-based imaging modality is an emerging noninvasive medical imaging approach exploring the dielectric property of biological tissue that shows great potential in breast cancer detection. Chapter 16 discusses a correlated microwave acoustic imaging modality and numerical simulation using finite-difference time-domain analysis. It is clearly shown that a combination of microwave-based imaging modalities is expected to provide an efficient diagnostic method for breast cancer detection in the future.

Fluorescence-based bioassays are novel diagnostic tools that are available to clinicians for deciding future treatment and to researchers for monitoring biological functions that may lead to novel investigations.







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The different aspects of photonic crystal fiber, its guiding mechanism, the refractive index law, etc. are analyzed and explained in Chapter 17. The proposed methodology is implemented in an array format of immuno recognition of specific proteins using a hollow-core photonic crystal fiber.

An overview of a quality-assurance program for digital mammography is discussed in Chapter 18. This overview includes the quality-control test procedures based on the American College of Radiology and the International Atomic Energy Agency. The role of medical physicists in the mammography quality-assurance programs, including acceptance, annual, and regular quality-control testing, is briefly presented.

In this book, we have made an honest effort to present information and methodologies for accurate diagnosis of breast cancer to help researchers, doctors, teachers, and students in biomedical science and engineering.

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Acronyms and Abbreviations

ABSN angles between surface normals
ABVS Automated Breast Volume Scanner

ACO ant colony optimization

ACR American College of Radiology

ACS ant colony system

AEC automatic exposure control

AI artificial intelligence

AJCC American Joint Committee on Cancer

ALA 5-aminolevulinic acid
ANN artificial neural network
ANOVA analysis of variance
APC antigen-presenting cell
APD avalanche photodiode

ART algebraic reconstruction technique

AUC area under ROC curve

AUCEC area under the contrast-enhancement curve

B+F base + fog BC breast cancer

BCDSG Breast Cancer Disease Site Group

BEM boundary element method

BFGS Broyden–Fletcher–Goldfarb–Shanno (method)

BHTE bioheat transfer equation

BIRADS[®] Breast Imaging Reporting and Data System

BIRADS-MRI BIRADS for MRI
bpp bits per pixel
BRCA-1, -2 breast cancer genes
BS bone scintigraphy
BSE breast self-examination

BWI bound water index
CaCo colorectal cancer

CAD computer-aided diagnosis
CADe computer-aided detection
CADx computer-aided diagnosis

CAM combined autocorrelation method

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CBE clinical breast examination
CCD charge-coupled device
CDS clinical decision support
CDSS clinical decision support system

CDU color Doppler ultrasound

CECT contrast-enhanced computed tomography

CFD computational fluid dynamics

CI confidence interval CLS curvilinear structure

CMAI correlated microwave acoustic imaging

CMM coordinate measuring machine

CMOS complementary metal-oxide semiconductor

CNS central nervous system CR computed radiography CR contrast response CTcomputed tomography **CTL** cytotoxic T lymphocyte **CTS** chaotic time series CW continuous wave **CXR** chest radiology DC dendritic cells

DCE-MRI dynamic contrast-enhanced MRI

DCIS ductal carcinoma in situ

DCS diffuse correlation spectroscopy

DD density difference

DDSM Digital Database for Screening Mammography

del detector element
DeTr decision tree

DICOM Digital Imaging and Communications in Medicine

DM digital mammography
DMD digital micromirror device

DMIST Digital Mammographic Imaging Screening Trial

DOI diffuse optical imaging
DOS diffuse optical spectroscopy
DOT diffuse optical tomography

DR digital radiology

DRS diffuse reflectance spectroscopy

DW diffusion weighted

E-M expectation maximization

EGFR epidermal growth factor receptor

EI exposure index EM electromagnetic

EORTC European Organization for Research and Treatment of Cancer







ER estrogen receptor

ESAK entrance surface air kerma **ESF** edge spread function EWevanescent wave fat-sat fat saturated **FCC** fibrocystic change FD frequency domain

fluorodeoxyglucose (18F) **FDG FDM** finite-difference method

fDOT fluorescence diffuse optical tomography

FDTD finite-difference time domain

FEM finite-element method fluoro-17β-estradiol FES

FFDM full-field digital mammography

FG fibroglandular FG fractal geometry

FLDA Fisher linear discriminant analysis

FN false negative

FNA fine-needle aspiration

FNAC fine-needle aspiration cytology

FNN false nearest neighbor

FP false positive **FROC** free-response ROC FS feature selection **FUI** feature usability index

GLCM gray-level co-occurrence matrix

GMM Gaussian mixture model

GSDF grayscale standard display function

GUI graphical user interface H&E hematoxylin-eosin **HCC** hepatocellular carcinoma

HC-PCF hollow-core photonic crystal fiber

HC/UFPE Clinical Hospital of the Federal University of

Pernambuco

HIFU high-intensity focused ultrasound

HL7 health level 7 **HNG** high nuclear grade

HRT hormone replacement therapy

HSP heat shock protein HVL half-value layer

IAEA International Atomic Energy Agency

ICG indocynanine-green **IDC** invasive ductal carcinoma







IFN interferon

IG-NIRS image-guided NIR spectroscopy

IM internal mammary

ING intermediate nuclear grade

IR infrared

ITK Insight Toolkit IV intravenous

JAFROC jackknife alternative free-response ROC

JND just-noticeable difference k-NN k-nearest neighbor

LA laser ablation

LBP local binary pattern

LC-VCO inductor/capacitor voltage-controlled oscillator

LCIS lobular carcinoma in situ
LE Lyapunov exponent
LNG low nuclear grade
LSM least-square method

LTB lesion-to-background (ratio)

LUT lookup table
MCC microcalcification
MD mid-density

MGD mean glandular dose

MGH Massachusetts General Hospital

MIAS Mammographic Image Analysis Society

MIP maximum-intensity projection

MIPT maximum-intensity projection over time mIPT mean-intensity projection over time

ML maximum likelihood

MOF microstructured optical fiber

MPV mean pixel value

MQSA Mammography Quality Standards Act

MR magnetic resonance

MRI magnetic resonance imaging MRS magnetic resonance spectroscopy

MTAI microwave-induced thermoacoustic imaging

MTF modulation transfer function

MWA microwave ablation MWI microwave imaging

NADH nicotinamide adenine dinucleotide plus hydrogen

NBC Naïve Bayesian classifier

NICE National Institute for Health and Clinical Excellence

(UK)

NIR near infrared







NPV negative predictive value

OI optical index

OO object-oriented (model)
ORD object-relational database

PASH pseudo-angiomatous stromal hyperplasia

PBG photonic bandgap

PBS phosphate-buffered saline PCF photonic crystal fiber PD progressive disease

PEM positron emission mammography
PET positron emission tomography
PET-CT PET with computed tomography
PDU power Doppler ultrasound

PDU power Doppler ultrasound
PHA phytohaemagglutinin
PMMA poly(methyl methacrylate)
PMT photomultiplier tube

PNN probabilistic neural network

PO pulse oximeter
PPIX protoporphyrin IX
PPV positive predictive value

PR partial response
PR progesterone receptor
QA quality assurance

QAP quality assurance program

QC quality control

QDA quadratic discriminant analysis

RBF radial basis function

RECIST response evaluation criteria in solid tumors

RF radiofrequency

RFA radiofrequency ablation

RI refractive index RI resistivity index

ROC receiver operating characteristic

ROI region of interest

ROL reference operating level RPS reconstructed phase space S/C signal-to-clutter ratio

SD stable disease

SDNR signal-difference-to-noise ratio SEM scanning electron microscope SFM screen-film mammography

SI shape index

SID source-to-image distance





SHH Sacred Heart Hospital

SIRT simultaneous iterative reconstruction technique

SLNB sentinel lymph node biopsy

SMPTE Society of Motion Picture and Television Engineers

SNR signal-to-noise ratio

SPAD single-photon avalanche diode

SPECT single-photon emission computed tomography

SQP sequential quadratic programming

std standard deviation

STIR short tau inversion recovery

StO₂ tissue oxygenation or oxygen saturation
SUVmax maximum standardized uptake value
SUVmean mean standardized uptake value
SVD singular value decomposition
SVM support vector machine
TBS tris-buffered saline
TBST tris-buffered saline/Tween

TD time domain

TDE time-delay embedding
TDLU terminal ductalobular unit
THC total haemoglobin concentration

THI tissue harmonic imaging
TIL tumor-infiltrating lymphocyte

TN true negative

TNF tumor necrosis factor

TP true positive

TPSF temporal point spread function

TS time series

TSFC texture shape feature coding
UCI University of California Irvine
US ultrasound/ultrasonography

US NCI United States National Cancer Institute
USPIO ultrasmall superparamagnetic iron oxide

UWB ultrawideband

VAB vacuum-assisted biopsy VOI volume of interest

WB whole body

WDBC Wisconsin Diagnostic Breast Cancer (database)
WFUSM Wake Forest University School of Medicine

WHO World Health Organization

WU Washington University in St. Louis



