

# DIELECTRIC MODELLING OF HUMAN SKIN AND BREAST TISSUE IN TERAHERTZ FREQUENCIES: POTENTIAL APPLICATION TO CANCER DETECTION

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## **Certificate of Authorship**

I certify that the work in this thesis has not previously been submitted for a degree nor has it been submitted as part of the requirements for a degree except as fully acknowledged within the text.

I also certify that the thesis has been written by me. Any help that I have received in my research work and the preparation of the thesis itself has been acknowledged. In addition, I certify that all information sources and literature used are indicated in the thesis.

Signature

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## Abstract

Growing developments in the generation and detection of terahertz (THz) radiation over more than two decades have created a strong incentive for researchers to study the biomedical applications of terahertz imaging. Contrasts in the THz images of various types of cancer, especially skin and breast cancer, are associated with changes in the dielectric properties of cancerous tissues. In fact, dielectric models can explain the interaction between terahertz radiation and human tissue at a molecular level just as their parameters have the potential for becoming indicators of cancer. However, dielectric modelling of various forms of human tissue remains limited due to a number of factors, especially suboptimal fitting algorithms and tissue heterogeneity.

Thanks to the high water content of human skin, its dielectric response to terahertz radiation can be described by the double Debye model. The existing fitting method using a nonlinear least square algorithm can extract the model parameters which track their measurements accurately at frequencies higher than one THz but poorly at lower frequencies. However, the majority of dielectric contrast between normal and cancerous skin tissues has been observed in the low THz range. Accordingly, this research has developed two global optimization algorithms which are capable of globally accurate tracking thereby supporting the full validity of the double Debye model in simulating the dielectric spectra of human skin in the THz frequencies. Numerical results confirm their superiority over the conventional methods. Furthermore, the next goal of the study is to apply statistical analysis to the parameters of the double Debye model in order to test their discrimination capability of skin cancer from normal tissue. Linear programming and support vector machine algorithms have also been employed using these parameters to classify normal skin tissue and basal cell carcinoma. By combining the double Debye parameters, the classification accuracy has shown significant improvement. The encouraging outcomes confirm the classification potential of the double Debye parameters.

The double Debye model, however, has been shown to be not suitable for simulating human breast tissue due to its low water content and heterogeneous structure, thus limiting the understanding of the THz dielectric response of breast tissue. To overcome this problem, this study proposes a new non-Debye dielectric model to fit the dielectric spectra of human breast tissue. Due to the mathematical complexity of the fitting procedure, a sampling gradient algorithm of non-smooth optimization is used to optimize the fitting solution. Simulation results confirm applicability of the non-Debye model through its exceptional ability to fit the examined data. Statistical measures have also been used to analyse the possibility of using the parameters of this model to differentiate breast tumours from healthy breast tissue. Based on the statistical analysis, popular classification methods such as support vector machines and Bayesian neural network have also been applied to examine these parameters and their combinations for breast cancer classification. The obtained classification accuracies indicate the classification potential of the model parameters as well as highlighting several valuable features of the parameter combinations.

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## Notation

THz NMSC	Terahertz Nonmelanoma Skin Cancer
BCC	Basal Cell Carcinoma
SCC	Squamous Cell Carcinoma
MMS	Moh's micrographic surgery
BCS	Breast conserving surgery
NS	Normal skin
$A \succ 0$	A is A positive semi-definite matrix
$A \ge 0$	A is positive, i.e all its entries are positive
$x \ge p$	$x_i \ge p_i$ with $x \in \mathbb{R}^n, p \in \mathbb{R}^n, i = 1, 2,, n$
$x \in [p,q]$	$x_i \in [p_i, q_i]$ with $x \in \mathbb{R}^n, p \in \mathbb{R}^n, p \in \mathbb{R}^n, i = 1, 2,, n$
$x \in \mathbb{R}^n_+$	$\{x \in \mathbb{R}^n : x_i \ge 0, i = 1, 2,, n\}$
$ ilde{x}$	x is a complex variable
$\Re(x)$	Real part of a complex variable $x$
$\Im(x)$	Imaginary part of a complex variable $x$
DD	Double Debye
THz-nDB	The non-Debye model for the human breast in the terahertz regime
NLS	Nonlinear Least Square
BB-GO	Global optimisation based on the branching and bounding method
BB-MNO	Minimax optimisation based on the branching and bounding method

QP	Quadratic program
$R^2_{adj}$	Adjusted $R^2$ indicates the goodness-of-fit
GOF	Goodness-of-fit
SVM	Support vector machine
HN	Havriliak-Negami relationship
CC	Cole-Cole equation
AVR	Average value
SE	Standard error
ROC	Receiver operating characteristic
AUROC	Area under the ROC curve
TPR	True positive rate
FPR	False positive rate
CV	Cross-validation
LOOCV	Leave-one-out cross-validation
RRSCV	Repeated random-subsampling cross-validation
RBF	Radial basic function
FDTD	Finite difference time domain
BNN	Bayesian neural network
NN	Neural network

## **Author's Publications**

The contents of this thesis are based on the following papers that have been published, accepted, or submitted to peer-reviewed journals and conferences.

#### Journal Papers:

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- Bao C. Q. Truong, Hoang D. Tuan, Anthony J. Fitzgerald, Vincent P. Wallace, and Hung T. Nguyen, "A dielectric model of human breast tissue in terahertz regime", *IEEE Transactions on Biomedical Engineering*, vol. 62, no. 2, pp. 699-707, Feb 2015.
- Bao C. Q. Truong, Hoang D. Tuan, Anthony J. Fitzgerald, Vincent P. Wallace, and Hung T. Nguyen, "The potential of the double Debye parameters to discriminate between basal cell carcinoma and normal skin", *IEEE Transactions on Terahertz Science and Technology*, vol. 5, no. 6, pp. 990-998, Nov 2015.

#### **Conference Papers:**

 Bao C. Q. Truong, Hoang D. Tuan, H. H. Kha, and Hung T. Nguyen, "System identification for terahertz wave's propagation and reflection in human skin", *Fourth International Conference on Communications and Electronics*, Hue, Vietnam, pp. 364-368, Aug 2012.

- Bao C. Q. Truong, Hoang D. Tuan, H. H. Kha, and Hung T. Nguyen, "Global optimization for human skin investigation in terahertz", 34th Annual International Conference of the IEEE Engineering in Medicine and Biology Society, San Diego, USA, pp. 5474-5477, Aug 2012.
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- 5. Bao C. Q. Truong, Hoang D. Tuan, Anthony J. Fitzgerald, Vincent P. Wallace, Tuan Nghia Nguyen, and Hung T. Nguyen, "Breast cancer classification using extracted parameters from a terahertz dielectric model of human breast tissue", 37th Annual International Conference of the IEEE Engineering in Medicine and Biology Society, Milan, Italy, pp. 2804-2807, Aug 2015.