

Volume 8. No. 3, March 2020

International Journal of Emerging Trends in Engineering Research Available Online at http://www.warse.org/IJETER/static/pdf/file/ijeter27832020.pdf

https://doi.org/10.30534/ijeter/2020/27832020

Diagnosing Osteoporosis through Numerical Simulation of Bone Sample by Non-Stationary Thermal Wave Imaging

Badugu Suresh, K Nikhel Sathvik, Syed Imran, O V Gopi Reddy, V Gopi Tilak, A Vijayalakshmi, V S Ghali

Infrared Imaging Center, Department of Electronics and Communication Engineering, Koneru Lakshmaiah Education Foundation, Vaddeswaram, AP, India

ABSTRACT

Infrared thermography is an emerging field of interest in diagnosis of various diseases. Osteoporosis is a skeletal disorder of reduced bone strength. A proper diagnosis of this disease further reduces the risk of bone fracture or further damages. In present paper, a bone sample of having different stages of osteoporosis is modeled using finite element analysis and simulated with a Quadratic frequency modulated stimulus. Further FFT phase, pulse compression and random projection transform techniques are employed as post processing schemes for better defect detection and compared their performance with signal to noise ratios.

Key words: Quadratic frequency modulated thermal wave imaging, Osteoporosis, Bone, infrared thermography, Pulse compression, FFT phase, Random projection transform.

1. INTRODUCTION

Osteoporosis is a skeletal disorder causing the reduced bone strength or density. As the age of a person grows, this may leads to risk of fracture. Though it cannot be reversed, early diagnosis helps to maintain good bone strength then avoid unbearable fractures. The nature of homeotherm possessed by human body and thermoregulation of inner core temperature will help to characterize diseases easily. Osteoporosis can be diagnosed based on the different thermal gradients produced in human skeletal structure due to different bone densities representing the stages of the disease. Though the blood perfusion is not considered, the changes in thermal parameters like density, specific heat and thermal conductivity of the diseased portion would be different from the sound region. Osteoporosis can be diagnosed through this basic principle and numerous authors presented their analysis on osteoporosis in biomedical point of view [1-4].

From the first demonstration of infrared thermography for diagnosis of physical illness by Barnes in 1963 [5], infrared thermography became an active research area in biomedical applications due to its safe, non-contact, non-invasive and wide area inspection characteristics. Application of infrared thermography went very pin point to further interests to characterize and diagnose different types of diseases based on infrared thermography. The extensive utilization of infrared thermography is identified in diabetic neuropathy [6], vascular disorder [7], breast cancer detection [8], fever screening [9], brain imaging [10], dry eye syndrome diagnosis [11], gynecology [12] and heart treatment [13]. Signal and image processing techniques incorporated with thermography will help to diagnose and analyze disease as they explore various properties of object under test.

Infrared thermography has its importance in industrial applications with the implementation of pulse [15, 16], lock-in [17], pulsed phase [18], frequency modulated [14] and quadratic frequency modulated thermal wave imaging techniques (FMTWI & QFMTWI) [19-31]. Recent advancement of frequency modulated thermal wave imaging on a bone sample with seven defects modeled using 3D finite element analysis and post processed using Fourier transform based phase analysis to retrieve the relation between relative bone density to relative phase [14]. In present work, the bone sample with different density variations is numerically modeled in COMSOL Multiphysics software using QFM stimulus and corresponding thermal variations are subjected to post processing schemes like FFT phase, pulse compression [20, 21] and random projection transform [22]. Corresponding qualitative analysis provided by thermographic visualization and defect signal to noise ratios.

2. QFMTWI

Non-stationary thermal wave imaging (NSTWI) techniques are gaining wide area of interest from past decade in infrared non-destructive testing (IRNDT). Frequency modulated thermal wave imaging (FMTWI) [14] and Quadratic frequency modulated thermal wave imaging (QFMTWI) [19] techniques comes under category of NSTWI. In QFMTWI [19], a selected band of low frequencies are supplied to modulate the optical stimulus which will heat up the top surface of the sample. The thermal wave generated along the top surface, propagates through the subsurface layers of sample and reflects from the termination end. These reflected gradients provide more heat on the surface. If any defects present in subsurface layers, that portion is more heated due to the time delays produced in the thermal gradients. This can be recorded through an infrared camera. The resultant thermal response is then given for post processing.

The general bio heat transfer equation considering all the arterial, Venus blood vessels, skin, fat and bone is that is widely used Penne's bio-heat transfer equation given by

$$\left(\rho c\right)_{\text{tissue}} \frac{\partial T}{\partial t} = k_{\text{tissue}} \frac{\partial^2 T}{\partial x^2} + \left(\rho c\right)_{\text{blood}} w \left(T_{\text{core}} - T\right) + Q_{\text{metabolism}}$$
(1)

Since the present work omits the considerations of blood perfusion w, temperature difference between core and arterial blood vessels, and the metabolism then the second and third terms in above equation tends to be zero. Then the bio-heat transfer equation modified to be general heat transfer equation with tissue thermal properties as

$$(\rho c)_{tissue} \frac{\partial T}{\partial t} = k_{tissue} \frac{\partial^2 T}{\partial x^2}$$
 (2)

Can be written as

$$\frac{\partial T}{\partial t} = \frac{1}{\infty} \frac{\partial^2 T}{\partial x^2} \tag{3}$$

Where $\alpha = (\rho c)$ tissue/k tissue is the diffusion coefficient, and T is temperature at depth x corresponding to the time instant t. Further Eq. [3] can be solved under boundary conditions with stimulating heat flux at skin surface, the obtained thermal response in Laplacian domain is given by

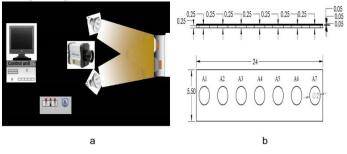
$$T(x,s) = \frac{Q(s).e^{-\sigma x}}{k\sigma}$$
(4)

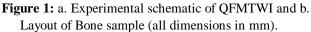
Where $\sigma = \sqrt{(s/\alpha)}$ and k is the thermal conductivity. Then the thermal response undergoes 1st order polynomial fitting to eliminate stationary component. Now, the retrieved dynamic thermal response is processed by FFT phase in which each thermal response is transformed into Fourier domain by applying FFT on it. The corresponding phase values are computed and respective phasegrams visualizes the thermal contrast between defective and non-defective regions.

Later a time domain cross correlation named pulse compression is analyzed to detect the time delays between defective and non-defective counterparts [20, 21]. This is employed by cross correlating each thermal response with a reference non-defective thermal profile. Similarly, random projection transform [22] also employed on thermal response which is a statistical method. In RPT, the 3D thermal response is reshaped into 2D and QR decomposition is applied over the 2D data to get orthonormal basis vectors. Further projecting few of these orthonormal basis vectors into data gives random projection components.

3. EXPERIMENTATION

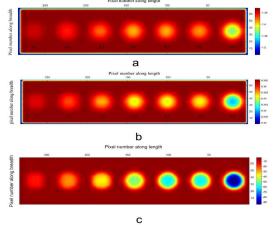
To test the proposed modality, a numerical simulation is carried out in COMSOL Multiphysics software using bio-heat transfer module. Bone sample consisting of skin, muscle and fat each of 0.5mm thickness and a bone of 2.5mm thickness with 7 holes having different density variations. The skin side of the sample excited by a QFM heat flux for 100 seconds with frequency range of 0.01Hz to 0.1Hz. Corresponding thermal response recorded at 25 frames per second. The experimental setup and layout of bone are shown in fig. 1. a and b respectively. Thermal properties of the sample are referred from[14-41].

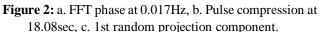




4. RESULTS AND DISCUSSION

The linear fitted and mean removed thermal response and Bartlett windowed thermal response is post processed through FFT phase; Pulse compression and Random projection transform techniques. The observed thermograms are given in fig. 2. a, b and c respectively. From figure, it is clear that the random projection transform distinguishes thermal variations of different bone densities efficiently.





Further the observed thermal response is characterized by performance metrics like signal to noise ratio. Signal to noise ratio is taken by dividing the difference between mean of defective region to mean of non-defective region by standard deviation of non-defective region[31-35,] as given below

$$SNR(dB) = 20\log\left(\frac{\mu_{Defective} - \mu_{Non-Defective}}{\sigma_{Non-defective}}\right)$$
(5)

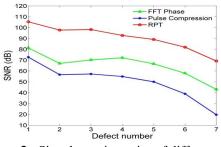


Figure 3: Signal to noise ratios of different post processing techniques employed on bone sample.

From the observations of thermograms and their respective defect signal to noise ratios, it is concluded that random projection transform provide better results for non-destructive evaluation of biomedical bone sample through QFMTWI.

5. CONCLUSION

The proposed work highlights the capabilities of quadratic frequency modulated thermal wave imaging for detecting density variations in simulated bone sample to characterize the severity of osteoporosis based on density variations. It is clear from the result that quadratic frequency modulated thermal wave imaging in combination with random projection transform based analysis can be used for the early, safe and remote diagnosis of osteoporosis with improved sensitivity and resolution. Along with detection, signal to noise ratios provide better results for random projection transform.

ACKNOWLEDGEMENT

This work was supported by FIST sponsored ECE Department under the grant no. **SR/FST/ET-II/2019/450**.

REFERENCES

- 1. Yaturu S, Humphrey S, Landry C and Jain S K. Decreased bone mineral density in men with metabolic syndrome alone and with type 2 diabetes, *Med. Sci. Monitor*, Vol. 15, no. 1, 2009.
- 2. Varenna M, Manara M, Galli L, Binelli L, Zucchi F and Sinigaglia L. The association between osteoporosis and hypertension: the role of a low dairy intake. *Calcified Tissue Int.* Vol. 93, no. 1, pp. 86–92, 2013. https://doi.org/10.1007/s00223-013-9731-9
- Hofbauer L C, Brueck C C, Shanahan C M, Schoppet M and Dobnig H. Vascular calcification and osteoporosis—from clinical observation towards molecular understanding. Osteoporosis Int., Vol. 18, no. 3, 251–9, 2007.
- Seo S K, Cho S, Kim H Y, Choi Y S, Park K H, Cho D J and Lee B S. Bone mineral density, arterial stiffness, and coronary atherosclerosis in healthy postmenopausal women. *Menopause*, Vol. 16, no. 5, 937–43, 2009.

- 5. R. B. Barnes. Thermography of the Human body. *Science*, Vol. 140, pp. 870-877, 1963.
- M. Bharara, J. E. Cobb and D. J. Claremount. Thermography and thermometry in the assessment of diabetic neuropathic foot: a case for further the role of thermal techniques. *International journal of lower extremity wounds*, Vol. 5, pp. 250-260, 2006.
- S. Bagavathiappan, T. Saravanan, J. Philip, T. Jayakumar, B. Raj, R. Karunanithi, T. Panicker, M.P. Korath and K. Jagadeesan. Infrared thermal imaging for detection of peripheral vascular disorders. *Journal of Medical Physics*, Vol. 34, pp. 43–47, 2009. https://doi.org/10.4103/0971-6203.48720
- 8. J.E. Thompson, T.L. Simpson and J.B. Caulfield. **Thermographic tumor detection enhancement using microwave heating.** *IEEE Transactions on Microwave Theory and Techniques*, MTT-26, 1978.
- 9. F. Ring and J. Mercer. **Thermal Imaging for fever** screening. *ISO Focus*, 2007.
- 10. I.A. Shevelev. Functional imaging of the brain by infrared radiation (Thermoencephaloscopy), *Progress in Neurobiology*, Vol. 56, pp. 269–305, 1998.
- B. Zelichowska, R. Rozycki, M. Tlustochowicz, A. Kujawa, B. Kalicki and P. Murawski. The usefulness of the thermography in the dry eye syndrome, *KlinikaOczna* Vol. 107, pp. 483–487, 2005.
- 12. C. Loriaux. Role of thermography in gynecology. Journal de Radiologie d Electrologieet de MedecineNucleaire, 56, 1975.
- A. Manginas, E. Andreanides, E. Leontiadis, P. Sfyrakis, T. Maounis, D. Degiannis, P. Alivizatos and D. Cokkinos, **Right ventricular endocardial** thermography in transplanted and coronary artery disease patients: first human application. *Journal of Invasive Cardiology*, Vol. 22, 2010.
- 14. Geetika Dua and Ravibabu Mulaveesala. Infrared thermography for detection and evaluation of bone density variations by non-stationary thermal wave imaging. *Biomed. Phys. Engg. Express*, Vol. 3, 2017. https://doi.org/10.1088/2057-1976/aa5b4d
- 15. Maldague X P V. Theory and Practice of Infrared Technology for Non-Destructive Testing. New York: Wiley, 2001.
- 16. Vladimir P. Vavilov and Douglas D.Burleigh. **Review of pulsed thermal NDT: Physical principles, theory and data processing.** *NDT&E International*, Vol. 73, 2015.
- 17. S. Pickering and D. Almond. Matched excitation energy comparison of the pulse and lock-in thermography NDE techniques. *NDT E Internat.*, Vol. 41, no. 7, 2008.
- PPT Ibarra-Castanedo C and Maldague X. Pulsed Phase Thermography Reviewed. Proceedings of QIRT 7 –Quantitative Infrared Thermography, 2004.
- 19. G. V. Subbarao and R. Mulaveesala. Quadratic frequency modulated thermal wave imaging for Nondestructive testing. *Progress in Electromagnetics Research M*, Vol. 26, pp. 11-22, 2012.

- Venkata Subbarao Ghali, Nataraj Jonnalagadda, and Ravibabu Mulaveesala. Three-Dimensional Pulse Compression for Infrared Nondestructive Testing. *IEEE Sensors Journal*, Vol. 9, no. 7, 2009.
- 21. Subhani S.K., Suresh B and Ghali V.S. **Empirical mode** decomposition approach for defect detection in non-stationary thermal wave imaging. *NDT and E International*, Vol. 81, pp. 39 - 45, 2016.
- S. Subhani, B.Suresh and V.S.Ghali. Orthonormal Projection approach for depth – resolvable subsurface analysis in non – stationary thermal wave imaging. *Insight*, Vol. 58, no. 1, pp. 42-45, 2016.
- 23. V. S Ghali, B.Suresh and A hemanth. Data fusion for enhanced defect Detectability in nonstationary thermal wave imaging. *IEEE sensors*, Vol. 15, no. 12, pp. 6761-6762, 2015. https://doi.org/10.1100/JSEN.2015.2472286

https://doi.org/10.1109/JSEN.2015.2472286

- 24. B Suresh, SK Subhani, A Vijayalakshmi, VH Vardhan and VS Ghali. Chirp Z transform based enhanced frequency resolution for depth resolvable non stationary thermal wave imaging. *Rev. Sci. Instr.*, Vol. 88, no. 1, 2017.
- B.Suresh, Sk. Subhani, V.S. Ghali and R. Mulaveesala. Subsurface details fusion for anomaly detection in non stationary thermal wave imaging. *Insight*, Vol. 59, no. 10, 2017.
- 26. Sk.Subhani, B.Suresh and V.S.Ghali. Quantitative subsurface analysis using frequency modulated thermal wave imaging. *Infrared Physics and Technology*, Vol. 88, 2018.
- 27. Md. M. Pasha, B.Suresh, K.RajeshBabu, Sk. Subhani and G.V.Subbarao. **Barker coded modulated thermal** wave imaging for defect detection of glass fiber reinforced plastic. *ARPN Journal of Engineering and Applied Sciences*, Vol. 13, no. 10, 2018.
- A. Vijaya Lakshmi, V. Gopitilak, Muzammil M. Parvez, S.K. Subhani and V.S. Ghali. Artificial neural networks based quantitative evaluation of subsurface anomalies in quadratic frequency modulated thermal wave imaging. *Infrared Physics & Technology*, Vol. 97, pp. 108-115, 2019.
- B.. Suresh, Jammula S.K and G.V Subbarao. Automatic detection of subsurface anomalies using non-linear chirped thermography. *International Journal of Innovative Technology and Exploring Engineering*, Vol. 8, no. 6, pp. 1247-1249, 2019.
- V.Krishna Chaitanya, N.Pranay manikanta, S.Mukesh Kumar, M.Anil Kumar, B.Suresh and V.S.Ghali. Defect Detection using Active Contour Method. *IJRTE*, Vol. 8, no. 4, pp. 10279-10282, 2019.
- Badugu Suresh. A Thermographic System for Enhanced Subsurface Detail Visualization through Automatic Detection. Ph.D Thesis, Department of Electronics and Communication Engineering, Koneru Lakshmaiah Education Foundation, Vaddeswaram, AP, India, 2019.

32. B. Suresh, M. Manorama, M. M. Bhupesh, K. Sai Kiran, G. V. P. Chandra Sekhar Yadav and V. S. Ghali. Advanced Signal Processing Approaches for Quadratic Frequency Modulated Thermal Wave Imaging. International Journal of Emerging Trends in Engineering Research (IJETER), Vol. 7, no. 11, pp. 599 – 603, 2019.

https://doi.org/10.30534/ijeter/2019/317112019

33. Badugu Suresh, T. Nikhilesh, T. Abhishek, M. Balakrishna, G. V. P. Chandra Sekhar Yadav and V. S.Ghali. Qualitative Subsurface Analysis in Quadratic Frequency Modulated Thermal WaveImaging. International Journal of Emerging Trends in Engineering Research (IJETER), Vol. 8, no. 1, pp. 31– 34, 2020.

https://doi.org/10.30534/ijeter/2020/06812020

- 34. B.Suresh, Sk.Subhani, S. Suparshya Babu, S. Susrutha babu. Matched Energy Modality for Non Stationary Thermal Wave Imaging. *IEEE international* conference on signal processing and communication engineering Systems Space, 2015.
- 35. R. Jaya Lakshmi, S. N. Sairam, G. Mounika, N.ayaram, V. Gopi Tilak, G. V. P. Chandra Sekhar Yadav and V. S. Ghali. Bartlett Windowed Quadratic Frequency Modulated Thermal Wave Imaging. International Journal of Emerging Trends in Engineering Research (IJETER), Vol. 7, no. 11, pp. 512-516, 2019. https://doi.org/10.30534/ijeter/2019/187112019
- 36. Sk Subhani, B. Suresh, K. Rajesh Babu and V.S. Ghali. Recent advancements in subsurface analysis with quadratic frequency modulated thermal wave imaging. *JATIT*, Vol. 95, no. 9, 2017.
- 37. T. S. V. Saketh, V. V. N. S. G. Sathwik, D. V. Sai Prahlad, V. Gopi Tilak, G. V. P. Chandra Sekhar Yadav and V. S. Ghali. Numerical Simulation of Non Destructive Characterization of CFRP Composite through Frequency Modulated Thermal wave Imaging, *IJRTE*, Vol. 8, no. 4, pp. 9754-9757, 2019.
- 38. B. Suresh, V. Pardha sardhi reddy, P S. Vinod and V S Ghali. Comparative Data Processing Methods for Non-Stationary Thermal Wave Imaging. *The first QIRT Asia conference on QIRT–ASIA*, 2015. https://doi.org/10.21611/qirt.2015.0105
- 39. B.Suresh, Sk Subhani and V.S. Ghali. Data Fusion Based Enhanced Defect Detection for Quadratic Frequency Modulated Thermal wave Imaging. National seminar NDE, 2016.
- 40. Sk. Subhani, B Suresh, Md Pasha, R Jayalakshamma and GV Subbarao. Chirp Z transform based Enhanced Defect Depth Resolution for Thermographic analysis of Composites. 26th NDE conference, 2016.
- B.Suresh, Sk. Shareena, Sk. Subhani and V.S.Ghali. Data Fusion for Subsurface Analysis in Non Stationary Thermal Wave Imaging. National seminar NDE, 2015.