

INTELLIGENT COMPUTING TECHNIQUES AND APPLICATIONS

Edited by

Tusharkanta Samal, Ambarish Panda, Manas Ranjan Kabat,
Ali Ismail Awad, Suvendra Kumar Jayasingh and Deepak K Tosh

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Preface

This Taylor & Francis, CRC Press volume contains the papers presented at the International Conference on Emerging Trends in Intelligent Computing Techniques (ICETICT – 2024) held during 27th and 28th December 2024 organized by DRIEMS University, Tangi, Cuttack, Odisha, India. A lot of challenges at us and no words of appreciation is enough for the organizing committee who could still pull it off successfully.

The conference draws the excellent technical keynote talk and many papers. The keynote talks by Prof. Sanjeevikumar Padmanaban, University of South-Eastern Norway and Prof. Bidyadhar Subudhi, Director, NIT, Warangal are worth mentioning. We are grateful to all the speakers for accepting our invitation and sparing their time to deliver the talks.

We received 238 full paper submissions and we accepted only 66 papers with an acceptance rate of 28%, which is considered very good in any standard. The contributing authors are from different parts of the globe that includes UAE, Philippines, Uganda, Jordan, Malaysia, and India. The conference also received papers from distinguished authors from the length and breadth of the country including 20 states and many premier institutes. All the papers are reviewed by at least three independent reviewers and in some cases by as many as five reviewers. All the papers are also checked for plagiarism and similarity score. It was really a tough job for us to select the best papers out of so many good papers for presentation in the conference. We had to do this unpleasant task, keeping the Taylor & Francis guidelines and approval conditions in view. We take this opportunity to thank all the authors for their excellent work and contributions and also the reviewers who have done an excellent job.

On behalf of the technical committee, we are thankful to Prof. Durga Prasad Mohapatra, NIT, Rourkela, to accept the invitation as General Chair of the Conference, for his timely and valuable advice. We also thankful to Prof. Srinibash Sethy for his constant support and guidance. We cannot imagine the conference without his active support at all the crossroads of decision-making process.

The university management of the host institute, the honorable Chairman Dr. Pramod Chandra Rath, Patron, Vice Chairman Durga Prasad Rath and Vice Chancellor, Prof. Prakash Kumar Hota, Director Admin Sj Balaram Kar, Director Coordination Mrs Chinmayee Rath, Co-Patron of the conference and Dean Prof. Nayan Ranjan Samal, Associate Dean Prof. Alok Ranjan Biswal, Organising Chair, Dean Research Prof. Biswaranjan Mohanty have extended all possible support for the smooth conduct of the Conference. Our sincere thanks to all of them.

We would also like to place on record our thanks to all the keynote speakers, reviewers, session chairs, authors, technical program committee members, various Chairs to handle finance, accommodation and publicity and above all to several volunteers. Our sincere thanks to all press, print and electronic media for their excellent coverage of this conference.

We are also thankful to Taylor & Francis, CRC Press publication house for agreeing to publish the accepted and presented papers.

Best wishes.
December, 2024

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1 An SRR metamaterial for focusing enhancement of W-band lens antenna for microwave imaging

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Abstract

This paper introduces a hybrid split ring resonator (SRR)-based periodic metamaterial structure developed for a W-band lens antenna, designed to detect electromagnetic radiation emitted by plasma or any cancerous cell. This detection enables localized analysis of the electron temperature from the radiation source. Here the metamaterial is tailored to operate within the W-band spectrum (70–110 GHz). The unit cell in the metamaterial structure is engineered for low loss, minimal dispersion, broad bandwidth, and high gain, optimizing lens performance. The structure's design, simulation, and optimization were conducted using ANSYS HFSS software. A composite metamaterial (CMM) was selected for the unit cell, achieving a near-zero refractive index to construct a low-loss, broadband, high-gain NZMTM (near-zero refractive index metamaterial) lens antenna that significantly enhances focusing. The lens aperture dimensions are 1100 μm × 550 μm × 250 μm.

Keywords: CMM, NZMTM, SRR, W-band

Introduction

Metamaterials have made significant strides in antenna systems, lens antenna, microwave detectors and filters that envisages for modern biomedical experiments for cancerous tissue imaging. Recently, lens antennas effectively focus electromagnetic energy into precise directions, achieving a narrow beam profile that maximizes signal strength in targeted areas as proposed by Abdul Mahmood [1]. Traditionally, dielectric materials are used as lens but the demerits of these materials as lens are their heaviness and bulkiness property, and their machining process is also expensive. Resonant and nonresonant sub wavelength periodic structures and artificial materials called metamaterial depicted by Bray et al. [2] are extensively used as lens for collimating the spherical wave into a focusing beam effectively for breast cancer detection. By choosing the specific sub wavelength metamaterial structure, we can control the effective material parameter. Thus, by manipulating the propagation of the EM wave, one can design different types of lens, e. g., negative index metamaterial lens (NIM), zero index metamaterial

lens (ZIM), and gradient refractive index metamaterial lens (GRIN) for bringing high resolution imaging in cancerous cell diagnosis proposed by Dixit and Pinto [3]. Metamaterials in microstrip patches are some specialized geometrical structures that do not exist naturally but are being designed artificially that depicts the negative or near zero value of permittivity, permeability and refractive index that enhances the tomographic microwave imaging quality as depicted by Grzegorczyk et.al, Ibrahim et.al, Islami et.al and Islam et.al [4-7]. Hence, these are also called left-handed materials (LHM) to enhance microwave imaging in biomedical instruments used particularly for cancer cell treatment proposed by Maamoun et al. [8]. This study introduces a multi-layered metamaterial design made up of unit cell components to develop a low-loss, high-efficiency, and broadband thin lens antenna. To achieve enhanced focusing and high-gain broadband performance, careful design and optimization of both the illuminator and unit cell are essential and suitable for biomedical imaging. These processes are conducted using ANSYS HFSS software to maximize performance.

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Unit Cell Design and Parameter Extraction

The proposed metamaterial unit cell, featuring split ring resonators (SRRs) positioned on one side of a cost-effective substrate, while a metallic strip is placed on the opposite side, directly aligned with the SRRs. The substrate, Rogers RT/Duroid 5880, has a relative permittivity of 2.0 and a loss tangent of $\tan \delta = 0.0009$. The SRR design incorporates two concentric rings, each with a strip width, separation, and split width measuring 55 μm . The outer ring has a diameter of 275 μm while the inner ring has a diameter of 110 μm . The substrate dimensions extend 550 μm along the x-axis and 1100 μm along the y-axis, with a thickness of 250 μm . The metallic strip on the unit cell measures 550 μm along the x-axis and 275 μm along the y-axis, with a copper layer thickness of 9 μm applied to both the SRRs and metallic strip. Simulations are conducted using a linearly polarized wave propagating in the Y-direction, with boundary conditions set for the E-field (PEC) in the X-direction and the H-field (PMC) in the Z-direction, as illustrated in Figure 1.1.

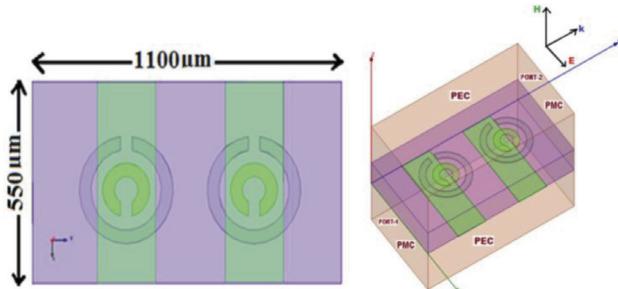


Figure 1.1 The proposed Unit Cell SRR structure and its simulation model with boundary conditions

Source: Author

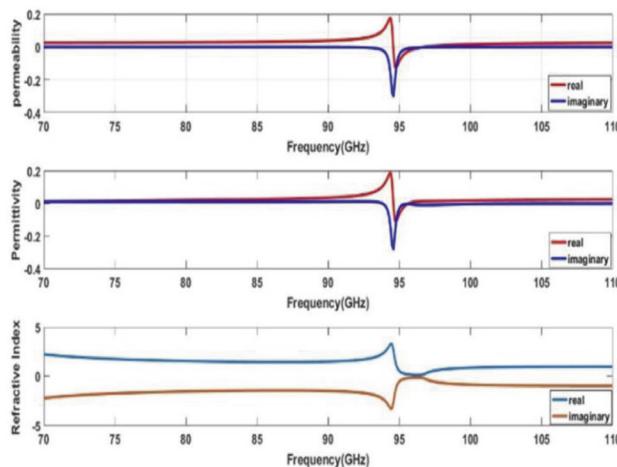


Figure 1.2 (a) Extracted medium parameters of the unit cell

Source: Author

The simulation utilizes ANSYS HFSS software, which operates on the principles of the finite element method (FEM) to analyze and optimize the design as considered by Afredi et.al.[9]. The effective medium parameters from the scattering parameter data are being extracted by the Nicolson–Ross–Weir (NRW) approach is shown in Figure 1.2. It is clear from the Figure 1.2(a) that the permittivity and permeability are near zero from 70-110 GHz with a resonance at 94 GHz that shows low electric and magnetic loss. In Figure 1.2(b), the dispersion diagram shows a phase (β) change of 1800 each from 87-92 GHz (LH band) and from 92-97 GHz (RH band) reveals near zero metamaterial behavior. In these ranges, the refractive index is almost near zero.

From the simulation of the unit cell structure, the transmission (S21) and reflection (S11), coefficients are shown in Figure 1.3(a). It shows that, in the band of 70-89 GHz, the S21 parameter shows lossy transmission and after that, S21 is apparent. The phase in angle degree for S11 and S21 is shown in Figure 1.3(b).

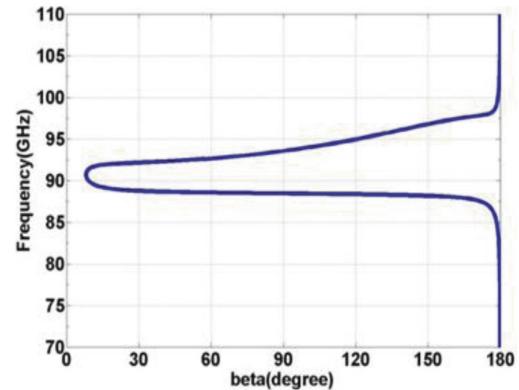


Figure 1.2 (b) Dispersion diagram for the proposed structure

Source: Author

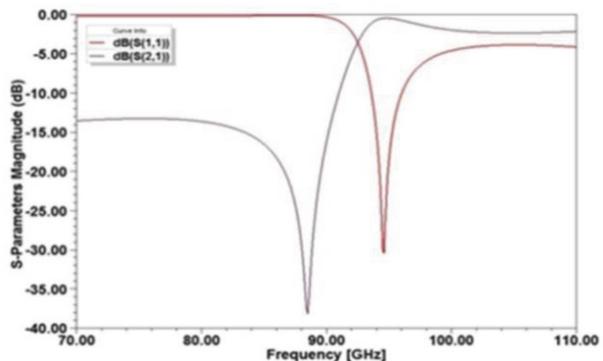


Figure 1.3 (a) Simulation results of magnitude of S11 and S21

Source: Author

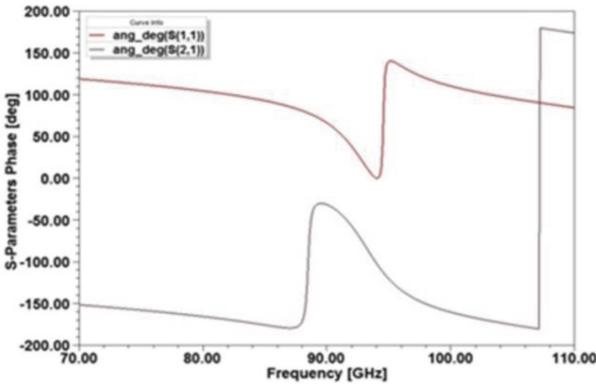


Figure 1.3 (b) Simulation result of phase of S11 and S21

Source: Author

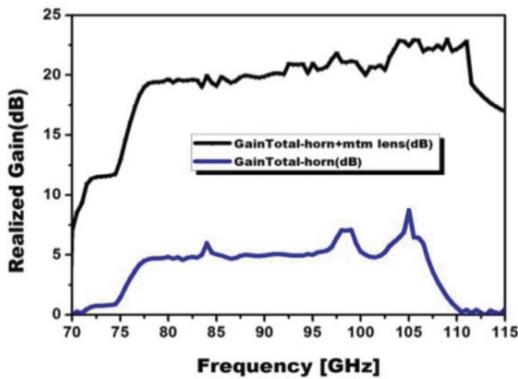


Figure 1.4 Maximum gain versus frequency plot
Source: Author

Lens Design and Simulation Result and Discussion

Metamaterial structure comprising of 10x10 unit cells in 10 layers is considered here to achieve high gain and directivity for a pyramidal horn in this W-band. The SRR lens-based horn is simulated for the desired frequency band and resulted gain versus frequency graph is depicted in Figure 1.4. It shows a consistency in gain enhancement of 10-12 dB in horn over a wide band of 77.5-102.5 GHz after the introduction of the parasitic SRRs.

In Figure 1.5, the 3D polar plot of gain of the lens antenna shows a gain of 22.053 dB in the propagation direction at 104.5 GHz. The lens antenna efficiently focuses electromagnetic waves in that direction, achieving a high concentration of signal strength compared to other directions. This would be beneficial in applications that require directed, high-intensity signal transmission, or reception. The gain is measured at a specific frequency, 104.5 GHz, which lies within the W-band (75–110 GHz). This implies that the antenna is designed or optimized to perform

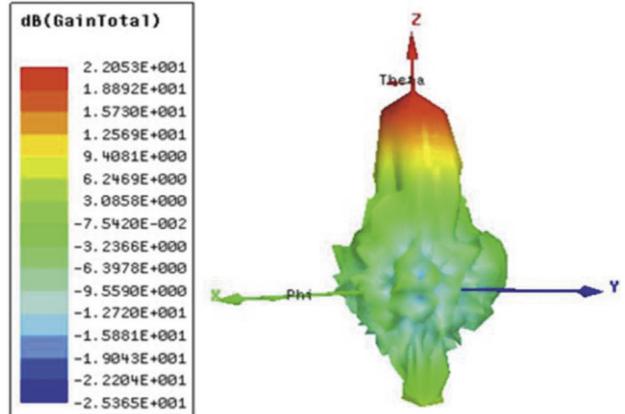


Figure 1.5 The 3D polar plot of gain at 104.5 GHz

Source: Author

particularly well at this frequency, making it suitable for applications in that band, such as radar, satellite communications, and high-frequency imaging systems.

Conclusion

This paper presents a periodic metamaterial structure based on a hybrid split ring resonator (SRR) designed for a W-band lens antenna aimed at absorbing and detecting electromagnetic radiation from plasma sources. The hybrid design incorporates double-negative (DNG) material properties, enabling it to act as an effective focusing lens with an innovative approach within this frequency range. Optimized for the W-band (70–110 GHz), this metamaterial lens holds potential applications in the development of wideband, high-gain antennas suitable for fusion plasma diagnostics, temperature measurement, radar, and imaging sensor systems.

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Hierarchically vectorized algorithmic paradigm for probabilistic node localization and pervasive threat detection in synthetically optimized WSN-assisted IoT architectures

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A genetic algorithm-optimized hybrid CNN-LSTM for robust EEG seizure detection and adversarial defense

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Utilization of deep learning and machine learning models to approach high glucose and low glucose prediction with type 1 diabetes mellitus in adult patients

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Modelling and performance analysis of a dexterous linkage-driven three-fingered under-actuated robotic hand

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Synthesis, structural and Raman study of Zn doped CuO nanoparticles

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Immersive Learning in Ayurveda: Revolutionizing Dhatus Poshan Nyaya education through AR/VR technologies

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Combating food insecurity through remote sensing and machine learning for enhanced crop yield prediction

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An assessment of vibration in non-prismatic Timoshenko beams with multiple transverse splits

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Robust digital image watermarking using hybrid optimization technique

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