



# 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 S. J. 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

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

## Editors Biography



**Dr. Tusharkanta Samal**  
Associate Professor  
DRIEMS University, Cuttack

Dr. Tusharkanta Samal holds a Bachelor of Technology in In received MTech and Ph.D. degree in Computer Science and engineering from Veer Surendra Sai University of Technology, Burla, Odisha, India. Currently he is working as an associate Professor in Department of Computer Science and engineering, DRIEMS University, Odisha, India. He has more than 12 years of teaching and research experience in different Engineering colleges and Universities. Dr. Samal has authored over 30 scientific papers published in SCI, Scopus Indexed Journals and Conferences.



**Dr. Ambarish Panda**  
Assistant Professor  
SUIIT, Burla

Dr. Ambarish Panda holds a Bachelor of Engineering in Electrical Engineering and Master of Technology in Power System Engineering from Sambalpur University and V.S.S University of Technology. He completed his Ph.D. in Electrical Engineering from V.S.S University of Technology since April, 2016. He has more than 14 years of teaching and research experience in different Engineering colleges and Universities. To his credit, he has published multiple research articles in SCI/SCIE indexed journals with high repute. Besides his research activities he serves as reviewer and Editor in SCI/SCIE Indexed Journal.



**Prof. (Dr.) Manas Ranjan Kabat**  
Professor and Principal  
IMIT, Cuttack

Manas Ranjan Kabat, is currently working as the principal of IMIT, a constituent college of BPUT. He has received his M.E. degree in Information Technology and Computer Engineering from Bengal Engineering College, India, and the Ph.D. degree in Computer Science and Engineering from Sambalpur University. He has more than two decades of teaching experience both at undergraduate and postgraduate level. He has published more than 75 research papers in various referred international journals and conferences. He has guided more than 20 M.Tech and 9 Ph.D. students. Chaired and organized many international and national conferences. Contributed as editor and reviewer of many peer-reviewed international journals. His research interests include QoS in internet, Wireless Sensor Network, Body Area Network, Cloud Computing etc.



**Dr. Ali Ismail Awad**  
Associate Professor, College of Information Technology, United Arab  
Emirates University (UAEU), Al Ain, United Arab Emirates

Ali Ismail Awad (Ph.D., SMIEEE, MACM) is an Associate Professor at the College of Information Technology, United Arab Emirates University (UAEU), Al Ain, United Arab Emirates, where he has been coordinating the master's program in Information Security since 2022. He is also an Honorary Associate Professor at the University of Nottingham, Nottingham, U.K. From 2018 to 2023, he served as a Visiting Researcher at the University of Plymouth, Plymouth, U.K. Dr. Awad joined the Department of Computer Science, Electrical and Space Engineering at Luleå University of Technology (LTU), Luleå, Sweden, in 2013. In 2017, he was promoted to Associate Professor (Docent) and served as the coordinator of the master's program in Information Security from 2017 to 2020. In recognition of his teaching merits and pedagogical achievements, he was promoted to the rank of Recognized University Teacher at LTU in 2021. His research interests include cybersecurity, network security, Internet of Things (IoT) security, and image analysis with biometrics and medical imaging applications. He has edited or co-edited several books and authored or co-authored numerous journal articles and conference papers in these areas of interest. From 2021 to 2024, he was recognized among the top 2% of influential scientists worldwide. Dr. Awad serves on the Editorial Boards of Future Generation Computer Systems, Computers & Security, Internet of Things; Engineering Cyber-Physical Human Systems, Health Information Science and Systems, and Security, Privacy and Authentication (Frontiers). Dr. Awad is an IEEE Senior Member and an ACM Professional Member.



**Dr. Suvendra Kumar Jayasingh**

Associate Professor and HOD in the Department of Computer Science and Engineering, Institute of Management and Information Technology (IMIT), Cuttack (A Constituent College of BPUT, Govt. of Odisha)

Dr. Suvendra Kumar Jayasingh is working as Associate Professor and HOD in the Department of Computer Science & Engineering, Institute of Management and Information Technology (IMIT), Cuttack (A Constituent College of BPUT, Govt. of Odisha) after being selected in OPSC (Orissa Public Service Commission) in 2005. He has obtained his Bachelor of Engineering the year 2003 from University College of Engineering (UCE), Burla (Now VSSUT). He got his M. Tech. in Computer Science & Engineering in 2007 from RVU, Udaipur and Ph. D. in Computer Science & Engineering in 2020 from North Orissa University, Baripada (Now Maharaja Sriram Chandra Bhanja Deo University). He is having 20 years of teaching experience in Computer Science & Engineering and MCA. He has published several articles, book chapters in reputed National and International journals and periodicals including Springer and Taylor & Francis and has presented research papers in National and International Seminars and Conferences. He has also participated in many National and International Workshops, FDPs, Industrial Training Programs organized by IITs, NITs and NITTTRs. His research interests include Artificial Intelligence, Data Mining, Soft Computing, Machine Learning, Computational Intelligence, Database Management System and Algorithm Analysis and Design. He has published a book on “Introduction to Machine Learning” and a UK patent on “Smart Home Air Quality Monitoring Device”. He is a life member of Indian Society for Technical Education (ISTE).



**Dr. Deepak K Tosh**

Assistant Professor, Department of Computer Science, University of Texas at El Paso, 500 W. University Avenue, El Paso, TX 79968-0518

Dr. Deepak K Tosh is an assistant professor in Computer Science at the University of Texas at El Paso. Before that he was a Cybersecurity Researcher at the DoD Sponsored Center of Excellence in Cybersecurity, Norfolk State University (NSU), Norfolk, Virginia. During that time, he has closely collaborated with researchers from Air Force Research Lab and Army Research Lab to establish data provenance mechanisms in cloud computing in addition to addressing research challenges in the arena of distributed system security, Blockchain, cyber-threat information sharing, cyber-insurance, and Internet of Battlefield Things (IoBT). He was appointed shortly as a postdoctoral researcher at the Tennessee State University, where he worked with Dr. Sachin Shetty in the

CyberViz Laboratory. He has been collaborating with Dr. Shetty since then focusing on research topics such as: distributed consensus models in Blockchain technology, cyber-resiliency in battlefield environment, and various practical issues in cloud computing security. He has received my Ph.D. in Computer Science and Engineering from University of Nevada, Reno, under the supervision of Dr. Shamik Sengupta. His dissertation was focused on designing market based models to enable cybersecurity information sharing among organizations. He received my masters in Computer Science from University of Hyderabad, India in 2012. His master thesis addressed the issue of cognitive radio transmission parameter adaptation problem using multi-objective optimization techniques.



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

*Debapriya Nilakantha Padhy<sup>1,a</sup>, Bajra Panjar Mishra<sup>2,b</sup>, Sudhakar Sahu<sup>3,c</sup>, Sakambari Mishra<sup>4,d</sup>, Subhranshu Kumar Singh<sup>2,e</sup> and Sudhanshu Bhusan Kar<sup>2,f</sup>*

<sup>1</sup>Department of Electronics and Communication, GIET, Bhubaneswar, Odisha, India

<sup>2</sup>School of Engineering and Technology, DRIEMS University, Cuttack, Odisha, India

<sup>3</sup>School of Electronics Engineering, KIIT University, Bhubaneswar, Odisha, India

<sup>4</sup>Department of Mathematics, College of Basic Science and Humanities, OUAT, Bhubaneswar, Odisha, India

## 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  $1100\ \mu\text{m} \times 550\ \mu\text{m} \times 250\ \mu\text{m}$ .

**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 Grzegorzczuk 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.

<sup>a</sup>hodece@gietbbsr.edu, <sup>b</sup>bajrapanjar@gmail.com, <sup>c</sup>ssahufet@kiit.ac.in, <sup>d</sup>sakambari@ouat.ac.in, <sup>e</sup>subhranshu@driems.ac.in, <sup>f</sup>director@driems.ac.in

## 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  $55 \mu\text{m}$ . The outer ring has a diameter of  $275 \mu\text{m}$  while the inner ring has a diameter of  $110 \mu\text{m}$ . The substrate dimensions extend  $550 \mu\text{m}$  along the x-axis and  $1100 \mu\text{m}$  along the y-axis, with a thickness of  $250 \mu\text{m}$ . The metallic strip on the unit cell measures  $550 \mu\text{m}$  along the x-axis and  $275 \mu\text{m}$  along the y-axis, with a copper layer thickness of  $9 \mu\text{m}$  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.

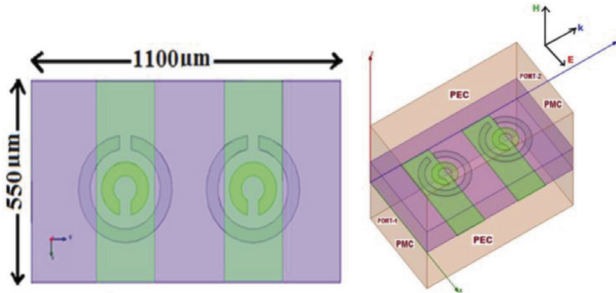


Figure 1.1 The proposed Unit Cell SRR structure and its simulation model with boundary conditions  
Source: Author

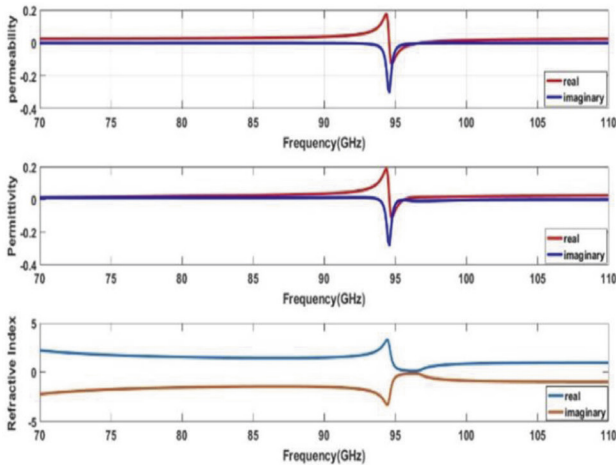


Figure 1.2 (a) Extracted medium parameters of the unit cell  
Source: Author

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 Afridi 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 permeability and permittivity 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 ( $\beta$ ) 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 ( $S_{21}$ ) and reflection ( $S_{11}$ ), coefficients are shown in Figure 1.3(a). It shows that, in the band of 70-89 GHz, the  $S_{21}$  parameter shows lossy transmission and after that,  $S_{21}$  is apparent. The phase in angle degree for  $S_{11}$  and  $S_{21}$  is shown in Figure 1.3(b).

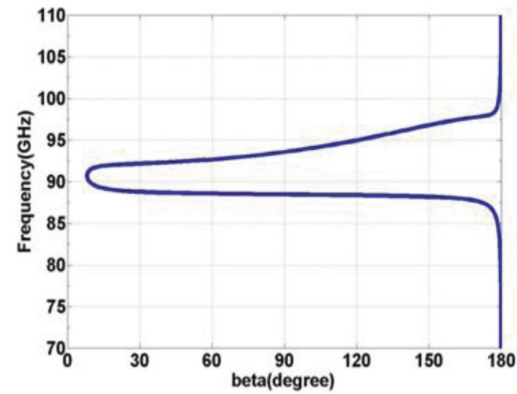


Figure 1.2 (b) Dispersion diagram for the proposed structure  
Source: Author

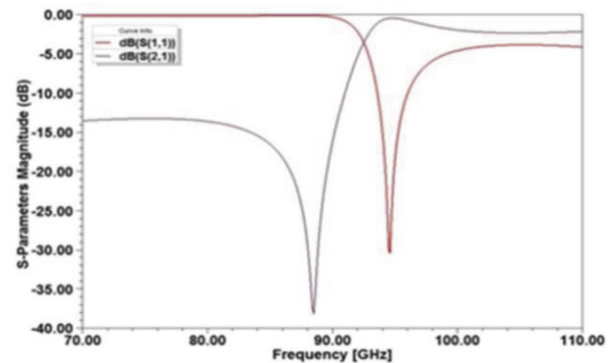


Figure 1.3 (a) Simulation results of magnitude of  $S_{11}$  and  $S_{21}$   
Source: Author

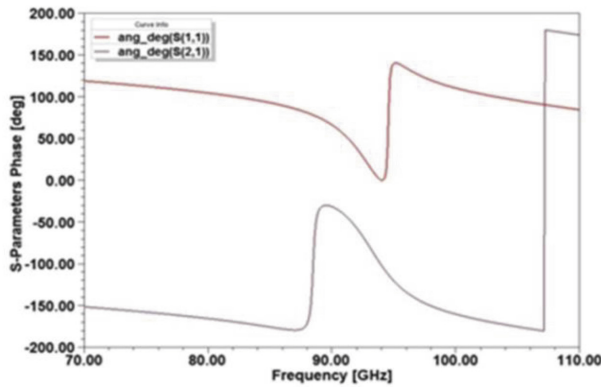


Figure 1.3 (b) Simulation result of phase of S11 and S21

Source: Author

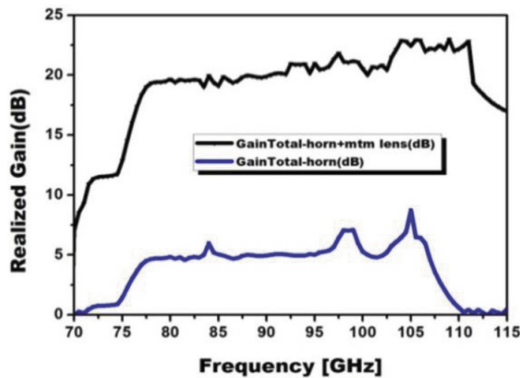


Figure 1.4 Maximum gain versus frequency plot

Source: Author

## Lens Design and Simulation Result and Discussion

Metamaterial structure comprising of  $10 \times 10$  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

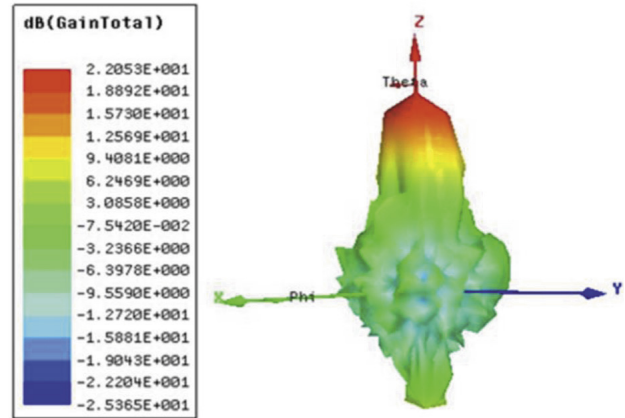


Figure 1.5 The 3D polar plot of gain at 104.5 GHz

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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## **A novel approach for mask face detection using convolutional neural network**

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## Intelligent strategies for the techno-economic evaluation of large-scale energy storage integrated hybrid power systems

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## Revolutionizing cancer diagnosis using machine learning techniques

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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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## Enhanced convolutional block attention decision Transnet for efficient IoT-driven patient monitoring and heart disease detection

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## Real time face mask detection in nuclear power plants: A deep learning framework using hybrid CNN-mobileNetV2 architecture

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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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## Predicting diabetic patients coronary artery calcium score, deep learning using retinal images

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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 Dhatu Poshan Nyaya education through AR/VR technologies

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## **GWO fuzzy PID controlled cuk and SEPIC converter based PFR**

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## **Enhancing web application security: A machine learning approach to firewall implementation**

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## **Combination of transition metal oxide into carbon nanofiber matrix for enhancement in electrochemical performance of supercapacitor**

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## **Experimental investigation on detection of medical image edges using radial basis fuzzy-neural network technique: A comparative analysis**

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## Hybrid vision transformer and CNN-based model for efficient cotton leaf disease detection

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## **Sensitivity analysis based performance assessment of a MGWO based PID control approach for regulating frequency in an interconnected power system**

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## **Real time garbage detection using CNN and YOLO algorithms**

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## Challenges and innovations in EV charging infrastructure: A focus on India's path to electrified transportation

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## **Nature-inspired metaheuristic algorithm optimized TIDN controller for frequency management in networked power system with storage device and HVDC link**

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## YOLO & ML based crack detection and strength prediction for structural health monitoring of bridges

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## Deep learning based optimized diagnostic model for diabetic retinopathy detection

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## **A novel U-Net with fine tuned VGG16 backbone model for ulcer detection and classification using 2-D endoscopy images**

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## Convolutional neural network approach to emotion recognition in speech

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## **A comparative analysis of breast cancer predictive intelligent model using machine learning techniques**

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## **An efficient CNN based hybrid model for cataract classification on ODIR dataset**

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

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## Electromagnetic interference shielding of Zn-50%-Al alloy-coated polypropylene flexible conducting film

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## **Robust digital image watermarking using hybrid optimization technique**

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