iJRASET
International Journal For Research in Applied Science and Engineering Technology

INTERNATIONAL JOURNAL FOR RESEARCH

IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Volume: 13 **Issue:** IX **Month of publication:** September 2025

DOI: https://doi.org/10.22214/ijraset.2025.73900

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International Journal for Research in Applied Science & Engineering Technology (IJRASET)
ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538
Volume 13 Issue IX Sep 2025- Available at www.ijraset.com

LLM's: The Progenitor of Future AI

Dr. Goldi Soni[1], Shubham Sharma[2], S. Satvika[3]
Department of Computer Science and Engineering, Amity University Chhattisgarh

Abstract: ***Large Language Models (LLMs) have transformed the landscape of artificial intelligence (AI) research, making significant contributions to areas such as natural language processing, scientific inquiry, and multimodal applications. This review examines the progression, functionalities, and obstacles related to LLMs. Recent developments in models such as GPT-4 and MiniGPT-4 have showcased substantial enhancements in reasoning capabilities, multimodal integration, and operational efficiency, facilitating applications that range from automated scientific exploration to interactive conversational agents. The review underscores the impact of LLMs across various sectors, including dentistry, where they improve diagnostic precision and treatment strategies, as well as in scientific research, where they aid in generating hypotheses and designing experiments. Furthermore, the emergence of 1-bit LLMs marks a significant shift towards cost-effective and energy-efficient AI solutions, enabling broader implementation. Nevertheless, challenges such as bias, ethical dilemmas, and high computational requirements persist. This paper synthesizes current research to offer a thorough overview of LLM evolution, applications, and prospective research avenues.***

Keywords: ***Large language models (LLMs), Multi modal learning, Deductive and inductive reasoning, Bit net architecture, LLM4SR.***

I. INTRODUCTION

The swift advancement of artificial intelligence (AI) has led to remarkable progress in natural language processing (NLP), with Large Language Models (LLMs) emerging as a groundbreaking technology across various sectors. Notably, models like OpenAI's GPT-4 exhibit outstanding abilities in text generation, logical reasoning, multimodal comprehension, and computational efficiency. LLMs have transformed areas such as healthcare, scientific inquiry, and computer vision, revealing new opportunities while also prompting discussions about ethical considerations, efficiency, and practical use. This paper offers a thorough examination of the latest developments in LLMs, emphasizing their applications in dentistry, multimodal AI, reasoning, efficient computational models, and the automation of scientific research.

II. EVOLUTION AND IMPACT OF LLMS

Large Language Models (LLMs) are constructed using deep learning frameworks, predominantly based on transformer architectures, which allow them to comprehend and produce text that resembles human language. The emergence of models such as GPT-3 and GPT-4 has greatly advanced Natural Language Processing (NLP), providing enhanced contextual awareness, creativity, and problem-solving capabilities. These improvements are primarily attributed to the increase in model parameters, the incorporation of multi-modal learning, and the application of reinforcement learning informed by human feedback.

A significant advancement in LLMs is their capacity to operate across various data types, including text, images, and audio. This versatility has facilitated their adoption in multiple sectors, such as healthcare, where they contribute to automated diagnostics, medical imaging, and patient management. The interdisciplinary influence of LLMs continues to grow, showcasing their potential to transform both professional and academic domains.

A. Applications of LLMs in Healthcare and Dentistry

The healthcare sector, including dentistry, is increasingly investigating the use of large language models (LLMs). A study examining ChatGPT's contribution to digital health underscores its potential for automating dental diagnosis and treatment planning (Satvika Reference 3). By utilizing multimodal AI, these models can interpret medical images, support clinical decision-making, and improve dental education. Additionally, AI-driven chatbots offer immediate assistance to both healthcare professionals and patients, thereby enhancing accessibility and efficiency in healthcare services.

However, the integration of LLMs into healthcare presents ethical and technical challenges. Concerns regarding data privacy, the accuracy of medical recommendations, and inherent biases in the models are significant issues that need to be addressed. Future research should aim to refine these models to improve their reliability and ensure adherence to medical regulations.

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B. Advancements in Multimodal AI and Vision-Language Models

The role of ions in dentistry is underscored by a study examining ChatGPT's contributions to digital health, particularly in automated dental diagnosis and treatment planning (Satvika Reference 3). By utilizing multimodal AI, these systems can effectively analyze medical images, support clinical decision-making, and improve dental education. AI-driven chatbots offer immediate assistance for both patients and practitioners. The fusion of vision and language models represents a significant leap in AI technology. MiniGPT-4 (Satvika Reference 2) exemplifies how the understanding of vision and language can be advanced through the use of sophisticated large language models (LLMs). By integrating a static visual encoder with a cutting-edge language model like Vicuna, MiniGPT-4 demonstrates multimodal capabilities akin to those of GPT-4, including the generation of detailed image descriptions and the interpretation of handwritten notes. These innovations have extensive applications, ranging from creating website designs based on sketches to providing real-time scene descriptions for individuals with visual impairments. The capacity of LLMs to process and integrate visual and textual data enhances their applicability across various sectors, including education, content development, and assistive technologies. Nonetheless, it is essential to address challenges such as dataset biases, inaccuracies in AI-generated content, and high computational expenses to effectively optimize these models for practical implementation.

C. Reasoning Capabilities in LLMs

One of the most fascinating features of large language models (LLMs) is their capacity to engage in reasoning tasks. A study on reasoning within LLMs (Satvika Reference 1) investigates how these models demonstrate cognitive-like reasoning abilities when presented with structured prompts. Methods such as Chain-of-Thought prompting allow LLMs to decompose intricate problems into sequential logical steps, thereby emulating human deductive reasoning.

Although LLMs exhibit competence in deductive, inductive, and abductive reasoning, their abilities still fall short when compared to human cognitive processes. The fundamental mechanisms that facilitate reasoning in LLMs are currently the focus of ongoing research. Gaining insights into how these models handle logical sequences, retain knowledge, and adjust to new information is essential for enhancing their decision-making precision. Assessing reasoning performance through comprehensive benchmarks will aid in pinpointing deficiencies and directing future advancements in AI-assisted problem-solving.

D. Efficient LLM Computation: The Rise of 1-bit Models

As large language models (LLMs) expand in size, their computational requirements have increased, raising concerns regarding energy consumption and deployment efficiency. Research on 1-bit LLMs (Shubham Reference 1) presents a novel approach to enhancing AI efficiency. By employing BitNet architectures with ternary weights (-1, 0, 1), these models achieve performance that rivals that of full-precision LLMs while markedly decreasing memory usage, latency, and energy expenses.

This advancement is especially significant for implementing AI in environments with limited resources. The ability to reduce model size without compromising accuracy allows for broader accessibility, ranging from mobile applications to embedded AI systems. Furthermore, optimizing LLMs for lower-bit representations supports sustainable AI initiatives by lessening the environmental footprint associated with large-scale deep learning models.

E. LLMs in Scientific Research and Knowledge Automation

LLMs are revolutionizing the landscape of scientific research by automating essential components of the research process. A study on LLMs in scientific inquiry (Evolution of Heuristics) highlights how these models aid in generating hypotheses, planning experiments, and producing automated scientific writing. AI-powered tools can support researchers in conducting literature reviews, composing manuscripts, and even facilitating peer review, thus enhancing the efficiency of academic workflows.

Nonetheless, dependence on LLMs for scientific exploration requires careful consideration. Challenges such as factual inaccuracies, excessive reliance on existing literature, and ethical implications of AI-generated content must be addressed diligently. Developing evaluation frameworks and incorporating human oversight into AI-assisted research tools will help maintain credibility and reliability in scientific endeavours.

III. LITERATURE REVIEW

Large Language Models (LLMs) have undergone significant advancements, showcasing impressive abilities in natural language processing tasks, primarily due to transformer architectures and extensive pretraining on large datasets (Minaee et al., 2024). A bibliometric analysis of LLM research conducted from 2017 to 2023 reveals a substantial increase in both publications and collaborations, indicating a growing interest from both academic and industrial sectors in their applications across diverse fields,

including medicine, engineering, and social sciences (Fan et al., 2023). Additionally, the expansion of LLMs has resulted in the emergence of capabilities such as in-context learning, instruction adherence, and reasoning, which were absent in smaller models, thereby establishing LLMs as crucial components in the quest for artificial general intelligence (Zhao et al., 2024). Nonetheless, challenges such as ethical issues, high computational costs, and the need for alignment with human values continue to be vital areas for future research.

Recent developments in Large Language Models (LLMs) have resulted in groundbreaking applications, including self-reflective reasoning, optimized memory management, and the evolution of algorithms. Abdali et al. (2025) propose a self-reflection technique based on the Hegelian dialectic, which allows LLMs to critically assess and enhance their outputs in a cyclical manner, thus fostering creativity and the synthesis of ideas. In a different approach, Kwon et al. (2023) tackle the computational inefficiencies associated with LLM deployment by introducing PagedAttention, an innovative memory management strategy that reduces waste in key-value cache storage, leading to improved throughput and lower operational expenses. Additionally, Liu et al. (2023) unveil Algorithm Evolution using Large Language Models (AEL), a method that automates the design of optimization algorithms within an evolutionary framework, thereby decreasing dependence on human expertise and boosting problem-solving capabilities. Collectively, these studies highlight the increasing adaptability of LLMs in enhancing reasoning, efficiency, and automation.

Large Language Models (LLMs) are playing an increasingly pivotal role in shaping scientific research, enhancing computational efficiency, and improving contextual accuracy. Luo et al. (2025) investigate the transformative impact of LLMs on the research process, facilitating hypothesis generation, experimental design, scientific writing, and peer review, thereby expediting scientific exploration. Concurrently, Kwon et al. (2023) tackle the computational challenges associated with LLM implementation by presenting PagedAttention, a memory management strategy that significantly mitigates fragmentation and boosts inference throughput by 2 to 4 times. In another study, Zhou et al. (2023) focus on the difficulties related to contextually accurate prompting, revealing that LLMs frequently rely on parametric knowledge rather than contextual cues, which can result in factual discrepancies; they suggest using counterfactual demonstrations and opinion-based prompts to enhance accuracy. Together, these studies underscore the growing significance of LLMs in fostering scientific innovation, efficiency, and dependability.

The security, learning potential, and exploratory capabilities of Large Language Models (LLMs) remain vital research domains. Abdali et al. (2024) outline significant security challenges that LLMs face, including adversarial attacks, privacy concerns, and the risk of model exploitation. They suggest various mitigation strategies, such as red teaming, model editing, and watermarking, to facilitate safer implementations. In a related study, Yang et al. (2024) reveal that integrating supervised knowledge from models fine-tuned for specific tasks greatly improves the in-context learning capabilities of LLMs, leading to a decrease in hallucinations and enhanced adaptability to out-of-distribution tasks. Additionally, Krishnamurthy et al. (2024) explore the capacity of LLMs to perform exploratory decision-making, noting that while models like GPT-4 can effectively explore when provided with well-structured prompts, they typically struggle without external summarization or fine-tuning. Collectively, these studies underscore the significance of security, guided learning, and structured prompting in enhancing the capabilities of LLMs.

Recent developments in Large Language Models (LLMs) have propelled advancements across multiple fields, including healthcare facilitation, model scalability, and the assessment of language models. Abdulhamid et al. (2024) investigate how LLMs can assist in managing medical chat groups, particularly for adolescents living with HIV in Kenya. Their findings emphasize the potential of LLMs to alleviate the challenges of moderation, while also raising ethical and practical issues related to accuracy and user engagement. In a different vein, Ma et al. (2024) present BitNet b1.58, an innovative 1.58-bit LLM architecture that significantly lowers computational expenses while delivering performance on par with full-precision models, thus establishing a new standard for efficient AI scaling. Furthermore, Watts et al. (2024) introduce PARIKSHA, a clear and scalable evaluation framework aimed at assessing Indic LLMs. This framework tackles the complexities of multilingual benchmarking by integrating both human and AI-driven evaluations. Together, these studies highlight the growing influence of LLMs in the realms of healthcare, efficiency, and evaluation practices.

The three research articles examine the development and effectiveness of large language models (LLMs) across different fields. Radford et al. (2019) emphasize the significance of unsupervised multitask learning in language models such as GPT-2, which exhibit impressive zero-shot transfer abilities across various NLP tasks without the need for specific fine-tuning. Touvron et al. (2023) present LLaMA, a series of open-source foundational models trained on publicly accessible datasets, revealing that smaller, well-optimized models can surpass larger proprietary models like GPT-3 in performance. Meyer et al. (2023) investigate the role of LLMs like ChatGPT in academic settings, highlighting their advantages in writing and research while also raising ethical concerns related to bias and the generation of inaccurate information. Collectively, these studies illustrate the progress made in LLMs, their influence on AI research, and the challenges associated with their ethical use.

The three research articles explore the advancement and efficacy of large language models (LLMs) in various domains. Radford et al. (2019) highlight the importance of unsupervised multitask learning in language models such as GPT-2, which demonstrate remarkable zero-shot transfer capabilities across multiple NLP tasks without requiring specific fine-tuning. Touvron et al. (2023) introduce LLaMA, a collection of open-source foundational models trained on publicly available datasets, showing that smaller, well-optimized models can outperform larger proprietary models like GPT-3 in terms of performance. Meyer et al. (2023) examine the impact of LLMs like ChatGPT in academic environments, underscoring their benefits in writing and research while also addressing ethical issues related to bias and the potential for generating misleading information. Together, these studies reflect the advancements in LLMs, their impact on AI research, and the ethical challenges that accompany their use.

The three research articles examine various dimensions of large language models (LLMs) and their role in education and reasoning. Huang and Chang (2023) present an extensive review of reasoning capabilities in LLMs, investigating methods such as chain-of-thought prompting, problem decomposition, and rationale engineering to improve the logical inference skills of AI. Bewersdorff et al. (2023) analyze the effectiveness of LLMs in evaluating student mistakes during scientific experiments, contrasting AI assessments with those of human evaluators. Their findings indicate that while AI excels at detecting basic errors, it encounters difficulties with more intricate issues. Katz et al. (2023) delve into the use of LLMs within engineering education research, illustrating how natural language processing (NLP) and generative AI can effectively analyze student essays, identify themes, and offer insights into students' career aspirations. Collectively, these studies underscore the increasing influence of LLMs in reasoning, education, and automated analysis, while also highlighting the necessity for enhanced accuracy and ethical considerations in the application of AI (Huang & Chang, 2023; Bewersdorff et al., 2023; Katz et al., 2023).

The three research articles examine various uses of large language models (LLMs) within the realms of healthcare and multimodal artificial intelligence. Zhu et al. (2023) present MiniGPT-4, a vision-language model that integrates a visual encoder with a sophisticated LLM, showcasing multimodal capabilities akin to GPT-4, including the generation of comprehensive image descriptions, the creation of websites from sketches, and storytelling based on images. Meskó and Topol (2023) advocate for the establishment of regulatory frameworks for LLMs in the healthcare sector, highlighting the importance of safety, ethical guidelines, and the protection of patient privacy, as these models play an increasingly significant role in clinical documentation, medical decision-making, and patient engagement. Reddy (2023) introduces a framework for assessing LLMs in healthcare, suggesting a translational value assessment model that incorporates governance mechanisms to ensure the accuracy, reliability, and ethical use of AI. Collectively, these studies underscore the transformative potential of LLMs in vision-language applications and healthcare, while also emphasizing the critical need for responsible implementation and regulation (Zhu et al., 2023; Meskó & Topol, 2023; Reddy, 2023).

The three research articles investigate the influence of large language models (LLMs) across various domains, concentrating on their uses, associated challenges, and ethical considerations. Ellaway and Tolsgaard (2023) analyze how LLMs affect academic authorship and scholarship within health professions education, raising questions about the ethical limits of AI-assisted writing, potential biases, and the accountability surrounding AI-generated materials. Huang et al. (2023) assess the capabilities of ChatGPT and multi-modal LLMs in the field of dentistry, highlighting their roles in automated dental diagnosis, treatment planning, and medical documentation, while also addressing issues related to data privacy, quality, and bias. Yan et al. (2023) perform a systematic review of LLMs in education, identifying 53 applications across various tasks such as grading, feedback, and content creation, while also noting challenges like transparency, replicability, and ethical issues concerning bias and fairness. Collectively, these studies illustrate the increasing impact of LLMs in education, healthcare, and dentistry, while emphasizing the importance of ethical considerations, transparency, and responsible application (Ellaway & Tolsgaard, 2023; Huang et al., 2023; Yan et al., 2023).

IV. COMPARISON OF RESEARCH PAPERS

Recent studies on Large Language Models (LLMs) collectively underscore their advancing capabilities in reasoning, multimodal comprehension, domain-specific uses, and computational efficiency. Progress includes the emergence of reasoning abilities, improved vision-language integration, potential applications in medical diagnostics, and the creation of low-bit architectures aimed at minimizing resource usage. These models exhibit potential in enhancing spatial reasoning, facilitating real-time interactions, and promoting overall AI scalability. Nevertheless, challenges remain, such as difficulties with intricate logic, hallucinations, issues related to data quality, and dependence on static training datasets. Collectively, these investigations highlight the necessity for a balanced approach to development that boosts performance while tackling concerns related to accuracy, adaptability, and ethical implementation.

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Table 4.1 Comparative analysis of five prominent papers

S.NO.	TITLE	AUTHOR	YEAR OF PUBLISH	OBJECTIVE	OUTCOME	FUTURE SCOPE	LIMITATION
1	Towards Reasoning in Large Language Models: A Survey.	Jie Huang	2023	Overview of reasoning in LLMs	Emergent reasoning abilities	Enhance reasoning capabilities of LLMs,	Struggle with complex reasoning
2	ENHANCING VISION-LANGUAGE UNDERSTANDING WITH ADVANCED LARGE LANGUAGE MODELS.	Deyao Zhu	2023	Investigate multimodal capabilities	Advanced vision language capabilities	Enhancing spatial reasoning	Hallucination and incorrect information
3	ChatGPT for shaping the future of dentistry: the potential of multi-modal large language model.	Hanyao Huang	2023	To shape the future of dentistry.	Potential of LLMs in dental diagonisis.	Real time patient interaction.	Data quality and accuracy issues.
4	The Era of 1-bit LLMs: All Large Language Models are in 1.58 Bits	Shuming Ma	2024	To examine evolution of LLMs.	Advancement in LLM capabilities.	Optimize computational efficiency.	Over-Reliance on Training Data and Lack of Real-Time Adaptation
5	Evolution of Heuristics: Towards Efficient Automatic Algorithm Design Using Large Language Model	Fei Liu	2024	To address ethical and practical challenges.	Automatic heuristics design (AHD) with LLMs	Real-World Deployment and Industry Applications	High Computational Cost and Energy Consumption

V. LIMITATIONS

Large Language Models (LLMs) represent significant progress in numerous fields; however, they also exhibit considerable limitations. A primary issue is the absence of clear logical frameworks in their reasoning processes, raising questions about whether these models genuinely "reason" or merely replicate established patterns. Problems related to the quality of datasets, the presence of noise, and inadequate alignment between different modalities contribute to unnatural language output and inconsistencies in multi-modal applications. In the realms of healthcare and dentistry, issues surrounding data privacy, biases inherent in training datasets, and the risk of producing inaccurate diagnoses impede their practical application in clinical settings. Moreover, while initiatives aimed at enhancing efficiency, such as lowering model precision to 1.58 bits, can improve cost-effectiveness, they may also jeopardize accuracy and necessitate specialized hardware for optimal functionality. In the context of scientific research, the capability of LLMs to formulate hypotheses and facilitate peer review raises questions regarding originality, validation, and ethical considerations. Furthermore, their smooth incorporation into existing workflows poses challenges across various disciplines.

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VI. FUTURE SCOPE

The future potential of Large Language Models (LLMs) encompasses various fields, with significant advancements anticipated. By enhancing reasoning capabilities through better logical structuring and optimized prompting methods, LLMs can achieve more dependable and transparent decision-making processes. In multi-modal applications, improving the synergy between visual and textual elements, along with utilizing high-quality datasets, will facilitate more fluid and precise language generation. In the realms of healthcare and dentistry, prioritizing data privacy, minimizing biases, and embedding LLMs into practical clinical workflows can bolster their diagnostic and treatment-support functions. Innovations aimed at efficiency, such as 1.58-bit models, have the potential to transform computational costs and energy usage; however, further investigation is necessary to ensure accuracy while enhancing hardware compatibility. In scientific research, employing LLMs for generating hypotheses, planning experiments, and conducting automated peer reviews could significantly expedite discoveries, provided that ethical considerations and validation issues are adequately addressed. Ultimately, ongoing improvements in model architectures, the ethical development of AI, and smooth integration into real-world applications will shape the forthcoming phase of LLM advancement.

VII. CONCLUSION

Large Language Models (LLMs) have shown significant promise in a variety of fields, including reasoning, vision-language comprehension, healthcare, scientific inquiry, and computational efficiency. However, despite their notable strengths, they face several challenges, including logical inconsistencies, data biases, privacy issues, and limitations related to hardware. The successful application of LLMs in practical settings, especially in healthcare and scientific research, demands further enhancements to guarantee accuracy, ethical standards, and user-friendliness. Innovations in model efficiency, such as low-bit precision architectures, present potential benefits in terms of cost and energy savings, yet require careful management to preserve performance levels. As research continues, it will be essential to tackle these challenges to fully realize the capabilities of LLMs, enhancing their reliability, interpretability, and applicability across various sectors.

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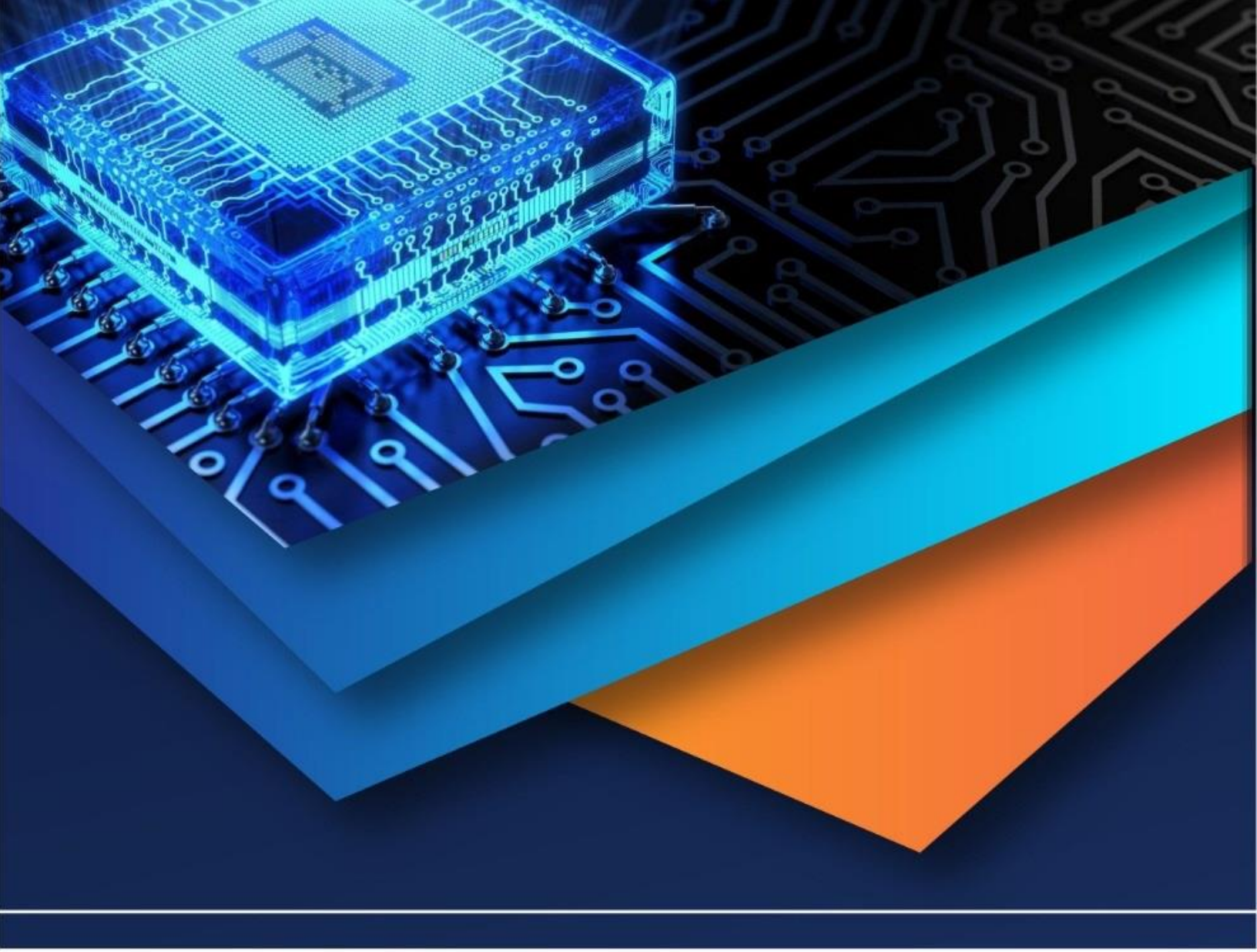

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