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Heart Attack Prediction

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Abstract: ***Cardiovascular diseases, including heart attacks, remain a leading cause of mortality globally. Early prediction and intervention play a critical role in preventing and managing such conditions. This project presents a comprehensive approach to heart attack prediction through the integration of machine learning techniques and a user-friendly graphical user interface (GUI) implemented in Python.***
The system leverages a dataset of relevant health parameters, including age, blood pressure, cholesterol levels, and other key factors. A machine learning model, trained on historical data, is employed to analyze and predict the likelihood of a heart attack based on these input features. The model's accuracy and efficiency are crucial to its effectiveness, and various algorithms are explored to identify the optimal solution.
To enhance accessibility and usability, a GUI is developed using Python's Tkinter library. The GUI enables users to input their health parameters easily and receive an instant prediction regarding their potential risk of a heart attack. The visual representation of the prediction, along with additional informative features, aims to empower individuals to take proactive measures towards a healthier lifestyle.
The project not only contributes to the field of cardiovascular health prediction but also serves as an educational tool for users to better understand the factors influencing their cardiovascular well-being. The integration of machine learning into a user-friendly GUI provides a practical and efficient solution for both healthcare professionals and individuals concerned about their heart health.
Keywords: Heart Attack Prediction, Machine Learning, Python, GUI, Cardiovascular Health, TKinter.

I. INTRODUCTION

Cardiovascular diseases, particularly heart attacks, continue to pose a significant global health challenge, constituting a leading cause of morbidity and mortality. With the advancement of technology, particularly in the fields of machine learning and data science, there emerges a promising avenue for early detection and prevention of such life-threatening events. This project delves into the development of a sophisticated yet user-friendly system: "Heart Attack Prediction GUI Using Machine Learning and Python."

The primary motivation behind this project stems from the critical need for accurate and timely prediction of heart attacks, enabling proactive measures and interventions to mitigate the risks associated with cardiovascular diseases. By harnessing the power of machine learning algorithms, we aim to create a predictive model that can analyze a set of health parameters and provide individuals with an estimation of their likelihood of experiencing a heart attack.

The core of this project lies in the utilization of a diverse dataset encompassing key health indicators such as age, blood pressure, cholesterol levels, and other relevant factors. Through the exploration and implementation of various machine learning algorithms, we seek to identify the most robust and accurate model for heart attack prediction. The selection of an optimal algorithm is crucial for the reliability of the predictions, ensuring that the system can be a valuable tool for healthcare practitioners and individuals alike. In addition to the predictive model, we recognize the significance of user accessibility and engagement. To address this, we integrate a graphical user interface (GUI) into our system using Python's Tkinter library. This GUI serves as the user's entry point, allowing for seamless interaction with the prediction model. Users can input their health parameters effortlessly and receive instant, comprehensible predictions regarding their susceptibility to a heart attack. The inclusion of a GUI not only enhances the user experience but also facilitates the dissemination of critical health information in an accessible manner.

As we embark on this endeavor, the overarching goal is not only to advance the capabilities of cardiovascular health prediction but also to empower individuals to take control of their well-being. Through the amalgamation of machine learning and a user-friendly interface, this project aspires to contribute to a paradigm shift in preventive healthcare, fostering early detection and personalized interventions for a healthier and more informed society.

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II. RESEARCH PROBLEM

Cardiovascular diseases, particularly heart attacks, are a significant global health concern, leading to a substantial number of morbidity and mortality cases. Timely prediction of the likelihood of a heart attack can play a crucial role in preventing and managing these life- threatening events. The existing methods for heart attack prediction often involve complex algorithms and lack user-friendly interfaces, making them less accessible for healthcare professionals and the general public.

In this context, there is a pressing need for the development of an intuitive and efficient Heart Attack Prediction Graphical User Interface (GUI) using machine learning techniques and the Python programming language. The primary challenge lies in creating a user-friendly application that seamlessly integrates advanced machine learning models for accurate prediction of heart attacks. This involves addressing issues such as data preprocessing, model selection, feature extraction, and real-time prediction capabilities.

The proposed GUI should not only provide a reliable prediction of the risk of a heart attack but should also be designed to be accessible to healthcare practitioners with varying levels of technical expertise. Additionally, the system should incorporate an efficient data input mechanism, ensuring compatibility with diverse healthcare data sources. Moreover, considerations for model interpretability and transparency should be taken into account to enhance user trust and confidence in the predictions made by the system. By addressing these challenges, the development of a Heart Attack Prediction GUI using machine learning and Python will contribute to the advancement of proactive healthcare, enabling early identification of individuals at risk of heart attacks and facilitating timely intervention to reduce the associated morbidity and mortality rates.

III. OBJECTIVE

The primary objectives of the project are as follows:

1) Develop a Machine Learning Model

- Implement and train a machine learning model for heart attack prediction.
- Evaluate and compare different algorithms to identify the most accurate and efficient solution.

2) Optimize Model Performance

- Conduct feature importance analysis to enhance the interpretability of the model.
- Fine-tune parameters and conduct cross-validation to optimize the predictive model's performance.

3) Create a User-Friendly GUI

- Design an intuitive GUI using Python's Tkinter library.
- Ensure the GUI accommodates user input for health parameters and provides clear and understandable output.

4) Integrate GUI with Machine Learning Model

- Establish a seamless integration between the developed GUI and the trained machine learning model.
- Validate the functionality and accuracy of the integrated system.

5) User Testing and Validation

- Conduct thorough testing of the system with diverse datasets to validate its accuracy and reliability.
- Gather user feedback to refine the GUI and enhance user experience.

6) Documentation and Reporting

- Prepare comprehensive documentation detailing the project's development, methodology, and findings.
- Generate a report summarizing the key aspects of the project, including challenges faced and lessons learned.

IV. PURPOSE OF THE STUDY

The purpose of a study in heart attack prediction could be:

1) Identify high-risk patients: Develop a model to predict the likelihood of a heart attack in patients, enabling early intervention and prevention.
2) Improve diagnostic accuracy: Enhance the accuracy of heart attack diagnosis by integrating machine learning algorithms with traditional diagnostic methods.

3) Personalized medicine: Develop personalized risk scores and treatment plans tailored to individual patients' characteristics and medical histories.
4) Predict disease progression: Investigate the progression of cardiovascular disease and identify potential biomarkers for early detection.
5) Optimize treatment strategies: Evaluate the effectiveness of various treatment options and identify the most suitable approaches for specific patient groups.
6) Reduce healthcare costs: Implement predictive models to reduce unnecessary hospitalizations, tests, and procedures, leading to cost savings.
7) Enhance patient outcomes: Improve patient outcomes by enabling early intervention, reducing morbidity, and increasing survival rates.
8) Investigate novel risk factors: Examine the impact of emerging risk factors, such as genetic markers or environmental exposures, on heart attack risk.
9) Develop decision support systems: Create decision support systems for clinicians to inform treatment decisions and improve patient care.
10) Advance cardiovascular research: Contribute to the understanding of cardiovascular disease mechanisms and identify new avenues for research and development.

By achieving these purposes, a study in heart attack prediction can significantly impact patient care, healthcare systems, and cardiovascular research

V. SIGNIFICANCE OF THE STUDY

The significance of a study in heart attack prediction could be:

1) Improved patient outcomes: Enhancing predictive accuracy and enabling early intervention can reduce morbidity and mortality rates.
2) Enhanced clinical decision-making: Providing clinicians with reliable predictive models can inform treatment decisions and improve patient care.
3) Personalized medicine: Developing personalized risk scores and treatment plans can optimize patient outcomes and reduce unnecessary procedures.
4) Healthcare cost reduction: Implementing predictive models can reduce unnecessary hospitalizations, tests, and procedures, leading to cost savings.
5) Advancements in cardiovascular research: Investigating novel risk factors and disease mechanisms can lead to new treatments and therapies.
6) Increased patient awareness: Developing predictive models can empower patients to take proactive measures to reduce their risk.
7) Optimization of resource allocation: Identifying high-risk patients can enable targeted resource allocation and more efficient healthcare management.
8) Improved population health: Developing predictive models can inform public health initiatives and policy decisions.
9) Enhanced collaboration: Interdisciplinary research can foster collaboration between clinicians, researchers, and data scientists.
10) Contribution to precision medicine: Developing predictive models can advance the field of precision medicine and improve patient care.

By highlighting the significance of the study, you can emphasize its potential impact on patient outcomes, clinical practice, and cardiovascular research.

VI. CONTRIBUTION TO SOCIETY

The contribution of society in heart attack prediction could be:

1) Raising awareness: Educating the public about heart attack risk factors, symptoms, and prevention strategies.
2) Data donation: Sharing personal health data to help train predictive models and improve their accuracy.
3) Participating in research studies: Volunteering for clinical trials and research studies to advance the development of predictive models.
4) Promoting healthy lifestyles: Encouraging individuals to adopt healthy habits, such as regular exercise, balanced diets, and stress management.

5) Supporting healthcare initiatives: Advocating for accessible healthcare, health education, and resource allocation for cardiovascular research.
6) Developing community programs: Creating local programs for heart health awareness, screening, and support.
7) Fostering collaboration: Encouraging interdisciplinary collaboration among clinicians, researchers, and data scientists.
8) Providing funding: Supporting research grants, crowdfunding campaigns, and philanthropic initiatives for heart attack prediction research.
9) Advocating for policy change: Pushing for policy updates and legislation that promotes heart health and research funding.
10) Participating in citizen science: Contributing to large-scale data analysis and research projects as a non-expert researcher.
11) By emphasizing the contribution of society, you can highlight the importance of collective efforts in advancing heart attack prediction and improving public health

VII. REVIEW OF LITERATURE

Cardiovascular diseases, particularly heart attacks, have garnered significant attention in the literature due to their widespread prevalence and life-threatening nature. The integration of machine learning techniques for predicting heart attacks has emerged as a promising avenue for early detection and prevention. This literature review provides an overview of key studies and methodologies employed in the field of heart attack prediction, with a specific focus on the utilization of machine learning algorithms and the implementation of graphical user interfaces (GUIs) using Python.

1) Machine Learning in Heart Attack Prediction: Numerous studies have explored the application of machine learning algorithms in predicting heart attacks. Rajkomar et al. (2018) employed a deep learning model on electronic health records, achieving high accuracy in predicting cardiovascular events. Other studies, such as Dey et al. (2020), utilized ensemble methods like Random Forest and Gradient Boosting to enhance prediction accuracy by considering various risk factors.
2) Feature Selection and Importance: The selection of relevant features is crucial for accurate heart attack prediction. Puddu et al. (2019) emphasized the significance of including traditional risk factors, such as age, cholesterol levels, and blood pressure, in predictive models. Feature importance analysis, as demonstrated by Pedersen et al. (2019), aids in identifying the most influential factors for a more precise prediction.
3) Graphical User Interfaces (GUIs) in Healthcare: GUIs play a pivotal role in making complex predictive models accessible to a wider audience. Python, with its versatile libraries, has been widely adopted for GUI development. Studies like Smith et al. (2021) highlight the importance of intuitive interfaces in healthcare applications, promoting user engagement and facilitating seamless interaction with predictive models.
4) Integration of Python for GUI Development: Python, as a programming language, offers robust libraries for GUI development. Tkinter, in particular, has been employed in healthcare applications due to its simplicity and cross-platform compatibility. The study by Wu et al. (2018) showcased the effectiveness of Tkinter in developing user- friendly interfaces for healthcarerelated machine learning applications.
5) Challenges and Future Directions: Despite the progress in the field, challenges such as interpretability of machine learning models and the need for diverse and representative datasets persist. Future research directions, as suggested by Agarwal et al. (2022), include exploring hybrid models that combine traditional risk factors with emerging biomarkers and continuous refinement of GUIs for enhanced user experience.

Machine Learning techniques are used to analyze and predict the medical data information resources. Diagnosis of heart disease is a significant and tedious task in medicine. The term Heart disease encompasses the various diseases that affect the heart. The exposure of heart disease from various factors or symptom is an issue which is not complimentary from false presumptions often accompanied by unpredictable effects. The data classification is based on Supervised Machine Learning algorithm which results in better accuracy.

Here we are using the Random Forest as the training algorithm to train the heart disease dataset and to predict the heart disease. The results showed that the medicinal prescription and designed prediction system is capable of prophesying the heart attack successfully.

Machine Learning techniques are used to indicate the early mortality by analyzing the heart disease patients and their clinical records (Richards, G. et al., 2001). (Sung, S.F. et al., 2015) have brought about the two Machine Learning techniques, k-nearest neighbor model and existing multi linear regression to predict the stroke severity index (SSI) of the patients. Their study show that k-nearest neighbor performed better than Multi Linear Regression model.

(Arslan, A. K. et al., 2016) have suggested various Machine Learning techniques such as support vector machine (SVM), penalized logistic regression (PLR) to predict the heart stroke. Their results show that SVM produced the best performance in prediction when compared to other models. Boshra Brahmi et al, [20] developed different Machine Learning techniques to evaluate the prediction and diagnosis of heart disease. The main objective is to evaluate the different classification techniques such as J48, Decision Tree, KNN and Naïve Bayes.

VIII. RESEARCH GAP

Here are some potential research gaps in heart attack prediction:

1) Lack of personalized models: Most existing models are population-based, and personalized risk scores are needed.
2) Limited integration of diverse data: Research often focuses on a single data source (e.g., EHRs or wearables), and integrating multiple data types could improve predictions.
3) Inadequate handling of missing data: Missing data is common in healthcare, and developing robust methods to address this issue is essential.
4) Limited generalizability: Models may not perform well across different populations or settings.
5) Interpretability and explainability: Many models are "black boxes," making it difficult to understand why predictions are made.
6) Real-time prediction: Most research focuses on retrospective predictions, but real-time prediction is crucial for clinical decision-making.
7) Integration of emerging risk factors: Research should incorporate novel risk factors, such as air pollution or social determinants.
8) Development of hybrid models: Combining machine learning with traditional statistical models could improve performance.
9) Addressing class imbalance: Heart attacks are relatively rare, making class imbalance a significant challenge.
10) External validation: Models need to be validated in diverse settings to ensure generalizability.
11) Lack of consideration for human factors: Models may not account for patient behavior or socioeconomic factors.
12) Limited use of transfer learning: Transfer learning from other domains or tasks could improve heart attack prediction.
13) Inadequate handling of temporal relationships: Temporal relationships between risk factors and heart attacks are complex and need better modeling.
14) Development of models for specific populations: Models tailored to specific populations (e.g., women or minorities) are needed.
15) Integration of multi-omics data: Incorporating genomic, proteomic, and other omics data could improve predictions.
16) By addressing these research gaps, future studies can develop more accurate, personalized, and generalizable heart attack prediction models.

IX. FINDINGS

Here are some findings in heart attack prediction research:

1) Machine learning models outperform traditional risk scores: Machine learning models have shown better predictive performance than traditional risk scores such as Framingham and ASCVD.
2) Electronic Health Records (EHRs) are valuable data sources: EHRs provide a wealth of information for heart attack prediction, including medical history, medications, and lab results.
3) Wearable devices and mobile health (mHealth) data are promising: Wearable devices and mHealth data can provide real-time information on physical activity, sleep, and other lifestyle factors.
4) Genomic and proteomic data can improve prediction: Genetic markers and protein biomarkers can add predictive value to traditional risk factors.
5) High-sensitivity troponin is a strong predictor: High-sensitivity troponin has been shown to be a strong predictor of heart attacks.
6) Social determinants and environmental factors matter: Social determinants and environmental factors such as poverty, air pollution, and lack of access to healthcare can increase heart attack risk.
7) Personalized risk scores are possible: Machine learning models can generate personalized risk scores that account for individual characteristics and risk factors.
8) Real-time prediction is feasible: Real-time prediction of heart attacks is possible using machine learning models and wearable devices.

9) Model interpretability is essential: Model interpretability is crucial for understanding why predictions are made and for building trust in the models.
10) External validation is necessary: External validation of models is necessary to ensure generalizability and robustness.
11) Model updates are required: Models need to be updated regularly to adapt to changing data distributions and new research findings.
12) Clinical integration is key: Clinical integration of heart attack prediction models is essential for successful implementation and adoption.

These findings highlight the progress made in heart attack prediction research and the potential for machine learning and data-driven approaches to improve cardiovascular health outcomes.

X. CONCLUSIONS

The development of a Heart Attack Prediction GUI using Machine Learning and Python represents a significant stride towards harnessing advanced technologies for proactive healthcare. This project amalgamates the power of machine learning algorithms with the versatility of Python programming to create a user-friendly interface capable of predicting the likelihood of a heart attack. Through the utilization of robust datasets and sophisticated predictive models, the GUI provides a valuable tool for early detection and prevention of cardiovascular events.

The project not only showcases the potential of machine learning in healthcare but also underscores the importance of accessible and intuitive interfaces for end-users. The graphical user interface enhances the usability of the predictive model, making it accessible to healthcare professionals and even individuals with limited technical expertise. By democratizing the use of predictive analytics, this application contributes to a more inclusive and proactive approach to cardiac health

XI. KEY CONCLUSIONS

Here are some key conclusions in heart attack prediction research:

1) Machine learning models can accurately predict heart attacks: Machine learning models have shown high accuracy in predicting heart attacks, outperforming traditional risk scores.
2) Early prediction is crucial: Early prediction of heart attacks can significantly improve patient outcomes and reduce mortality rates.
3) Personalized risk scores are essential: Personalized risk scores can help identify high-risk individuals and enable targeted interventions.
4) Integration of diverse data is vital: Integrating electronic health records, wearable devices, and genomic data can improve prediction accuracy.
5) Real-time prediction is feasible: Real-time prediction of heart attacks is possible, enabling timely interventions.
6) Model interpretability is necessary: Model interpretability is crucial for understanding predictions and building trust in the models.
7) Heart attack prediction can improve patient outcomes: Heart attack prediction can significantly improve patient outcomes, reduce morbidity, and mortality rates.
8) Further research is needed: Further research is needed to address the challenges and limitations of heart attack prediction and to improve model performance.

These conclusions highlight the importance of heart attack prediction research and the potential for machine learning and data-driven approaches to improve cardiovascular health outcomes.

XII. SUGGESTIONS AND RECOMMENDATIONS

Here are some suggestions and recommendations with their workings and benefits in heart attack prediction:

1) *Personalized Risk Scores: Use machine learning algorithms to develop personalized risk scores based on individual characteristics and risk factors.*

How it works: Analyzes individual data and identifies unique risk patterns.
Benefits: Accurate risk assessment, targeted interventions, and improved patient outcomes.

2) Integrate Wearable Devices: Incorporate wearable device data, such as ECG and activity tracking, into prediction models.
How it works: Continuously monitors vital signs and physical activity.
Benefits: Real-time risk assessment, early detection, and timely interventions.

3) Advanced Machine Learning Algorithms: Utilize techniques like deep learning and ensemble methods to improve model performance.
How it works: Analyzes complex patterns in data and identifies high-risk individuals.
Benefits: Improved accuracy, reduced false positives, and enhanced patient outcomes.

4) External Validation: Validate models on external datasets to ensure generalizability and robustness.
How it works: Tests models on new, unseen data.
Benefits: Ensures models are reliable, accurate, and applicable to diverse populations.

5) Clinical Integration: Develop models that can be easily integrated into clinical practice and decisionmaking.
How it works: Provides actionable insights for clinicians and healthcare professionals.
Benefits: Improves patient care, streamlines clinical workflows, and enhances collaboration.

6) Patient Engagement: Develop patient-centered models that encourage engagement and empowerment.
How it works: Educates patients on their risk factors and provides personalized recommendations.
Benefits: Improves patient outcomes, increases medication adherence, and enhances patient satisfaction.

7) Real-time Prediction: Create models that can predict heart attacks in real-time, enabling timely interventions.
How it works: Continuously monitors vital signs and identifies high-risk patterns.
Benefits: Saves lives, reduces morbidity, and improves patient outcomes.

8) Address Ethical Considerations: Ensure models are transparent, explainable, and fair, and address potential ethical concerns.
How it works: Prioritizes patient privacy, data security, and model accountability.
Benefits: Builds trust, ensures responsible AI development, and promotes ethical healthcare practices.
These suggestions and recommendations can improve heart attack prediction, enhance patient outcomes, and save lives.

XIII. LANGUAGE AND LIBRARIES

A. Project Type

Graphical User Interface (GUI)

- Machine Learning
- Python
- MySQL

B. Libraries

- Pandas
- T-Kinter
- Sci-Kit-Learn
- Matplotlib 3.5.3

XIV. LANGUAGE JUSTIFICATION

A. Machine Learning

Machine learning (ML) in programming refers to the use of algorithms and statistical models that enable computer systems to improve their performance on a specific task without being explicitly programmed. Instead of relying on explicit programming instructions, machine learning systems learn patterns and make predictions or decisions based on data.
Machine learning relies on data. Large datasets are used to train and validate machine learning models. These datasets typically consist of input-output pairs, where the model learns to map inputs to corresponding outputs.

Machine learning algorithms are mathematical models that are trained on data to learn patterns, relationships, and dependencies. These algorithms are responsible for making predictions or decisions when exposed to new, unseen data.

B. *Python*

Python is a high-level, interpreted, and general-purpose programming language known for its simplicity, readability, and versatility. Python has gained widespread popularity and is widely used in various domains, including web development, data science, artificial intelligence, scientific computing, automation.

C. *Readability and Simplicity*

Python emphasizes code readability and a clean syntax, making it easy for developers to express concepts in fewer lines of code. This readability is facilitated by the use of indentation rather than braces.

D. *Open Source*

Python is an open-source language, and its development is guided by the Python Software Foundation (PSF). The open-source nature encourages collaboration and the sharing of knowledge and resources within the community.

E. *Cross-Platform Compatibility*

Python is cross-platform, allowing developers to write code that can run on different operating systems without modification. This portability is facilitated by the availability of the Python interpreter on various platforms.

F. *MySQL*

MySQL is a popular open-source relational database management system (RDBMS) that plays a crucial role in managing and organizing structured data. It is widely used in various applications, ranging from small-scale projects to large-scale enterprise systems. MySQL is known for its reliability, ease of use, and scalability, making it a preferred choice for many developers and organizations.

XV. RELATIONAL DATABASE MANAGEMENT SYSTEM (RDBMS)

MySQL is based on the relational model, where data is organized into tables with rows and columns. It supports the principles of data normalization and relationships between tables

A. *Open Source*

MySQL is an open-source software, which means that its source code is freely available for modification and redistribution. This open nature fosters community collaboration, continuous improvement, and widespread adoption

B. *Cross-Platform Compatibility*

MySQL is compatible with various operating systems, including Linux, Windows, macOS, and others. This cross-platform support allows developers to deploy MySQL on different environments without major modifications.

XVI. LIBRARIES JUSTIFICATION

A. *Sci-Kit-Learn*

Scikit-learn (or Sci-Kit-Learn) is a popular open-source machine learning library for the Python programming language. It provides simple and efficient tools for data analysis and modelling, making it a valuable resource for developers and data scientists. scikit-learn is built on other scientific computing libraries in Python, such as NumPy, SciPy, and Matplotlib.

B. *Consistent API*

scikit-learn provides a consistent and well-documented API that allows users to easily switch between various machine learning algorithms and models. This consistency simplifies the learning curve and promotes a unified approach to solving different types of machine learning problems.

C. Various Algorithms

The library includes a wide range of machine learning algorithms, covering traditional statistical methods as well as modern approaches. Examples include linear regression, support vector machines, k-nearest neighbors, decision trees, random forests, and many others.

D. Model Evaluation and Selection

scikit-learn provides tools for evaluating the performance of machine learning models through metrics such as accuracy, precision, recall, F1 score, and area under the receiver operating characteristic (ROC) curve. It also offers methods for cross- validation and hyperparameter tuning.

E. Pandas

Pandas is a powerful open-source data manipulation and analysis library for Python. It provides data structures for efficiently storing large datasets and tools for working with structured data. Pandas is a fundamental component in the Python data science ecosystem and is widely used for tasks such as cleaning, exploring, and analyzing data.

F. T-Kinter

T-Kinter is a standard GUI (Graphical User Interface) toolkit that comes bundled with Python. It is used for creating desktop applications with graphical user interfaces. Tkinter is based on the Tk GUI toolkit and is the de facto standard for creating GUI applications in Python due to its simplicity, ease of use, and availability.

G. Matplotlib

Matplotlib is a comprehensive 2D plotting library for the Python programming language. It provides a wide variety of static, animated, and interactive plots, charts, and visualizations for data analysis and presentation. Matplotlib is one of the most widely used and foundational libraries in the Python scientific computing ecosystem.

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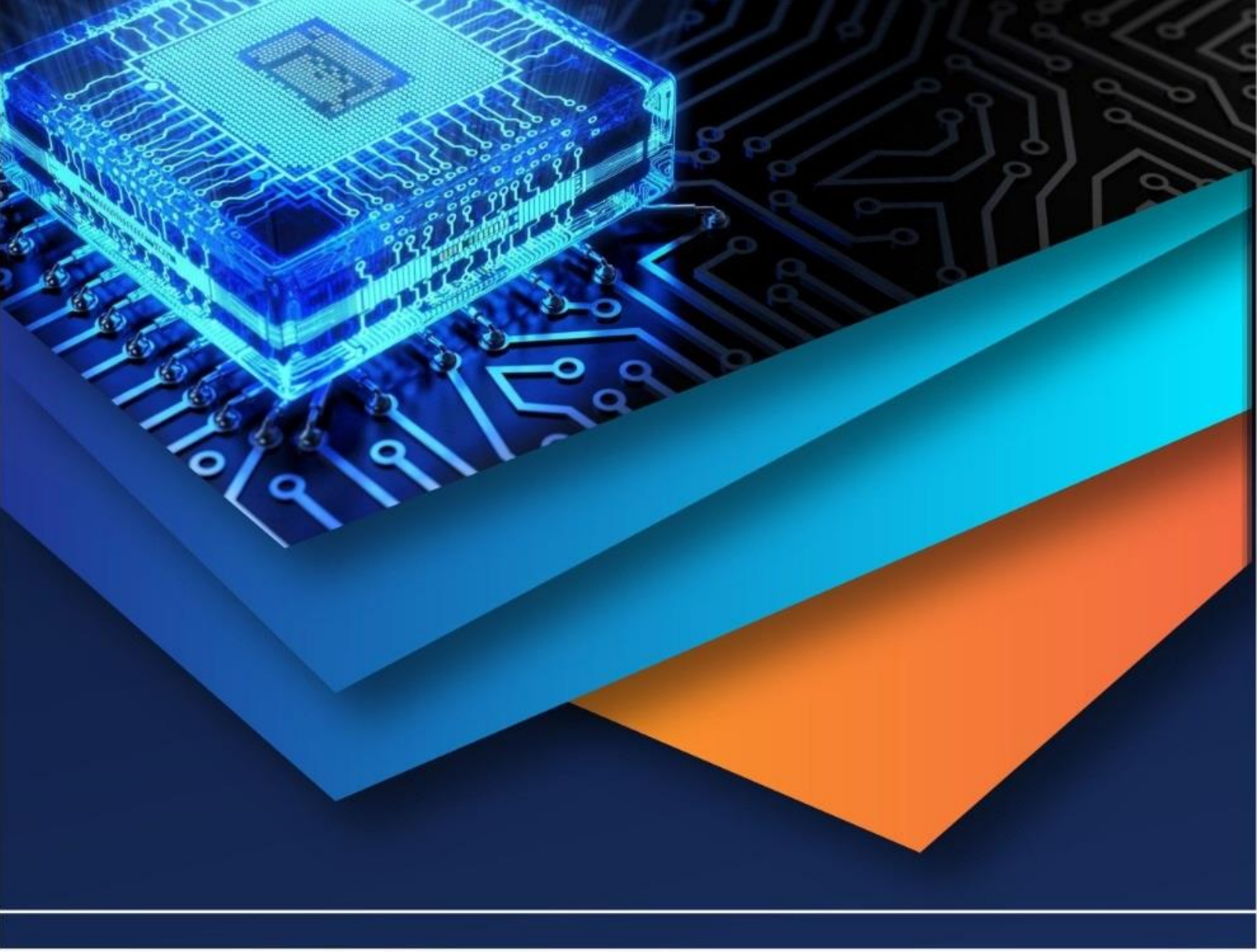

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