



Advanced Machine Learning for Early Detection and Alerting of Cardiac Stroke Events

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Abstract: Machine learning enabled cardiac stroke alert system uses temperature, heart rate, and machine learning algorithms to detect individuals who are at risk and those with a history of heart attacks. Real-time monitoring is made available by integrating IoT technology, providing medical personnel a way to respond quickly to alerts and take necessary steps. Wearable sensor-derived temperature and heart rate data is thoroughly examined by machine learning models that were trained on appropriate datasets. The objective of the new approach is to improve patient outcomes by enabling early diagnosis of cardiac events. The project provides an extensive system for continuous and remote patient monitoring, in addition to its focus on heart attack detection, through the integration of sensor technologies, machine learning, and Internet of Things connectivity with all factors taken into account, the Heart Stroke Alert System is a major step advance for cardiac care because it provides continuous patient monitoring and a preventive approach to event recognition. This system has the potential to change the detection and management of cardiac events by utilizing the synergy of sensor technologies, machine learning, and IoT connection. This could ultimately result in improved patient results and higher quality care.

Keywords: Heart Stroke detection, IOT, Machine learning, Real time monitoring, Early detection, Wearable sensor, Continuous patient monitoring.

1. INTRODUCTION

Heart strokes and cardiovascular diseases are significant contributors to global mortality rates, emphasizing the critical need for early identification and intervention. This research proposes an innovative approach to address this challenge by combining machine learning techniques, wearable sensors, and Internet of Things (IoT) technology. The device utilizes a temperature sensor and a heart rate sensor to collect physiological data, focusing on key indicators related to heart health. Machine learning techniques play a crucial role in analyzing the collected datasets, distinguishing between individuals who have previously experienced a heart stroke and those who may be at risk. By examining patterns and correlations within the data, machine learning algorithms can identify subtle changes in physiological parameters that may indicate an increased risk of cardiovascular events. Integration of IoT technology into the system framework enhances its capabilities for real-time monitoring and response. This connectivity enables seamless communication between the device, medical staff, and healthcare infrastructure, facilitating immediate emergency

response when necessary.

Regular interaction between patients and healthcare providers is enabled, allowing for ongoing monitoring and timely intervention. The initiative aims to offer an integrated approach for ongoing patient monitoring along with early stroke detection. By combining developments in sensor technology, machine learning, and IoT connectivity, the project sets up a proactive healthcare system that enables individuals to effectively manage their heart health. This holistic approach empowers patients to take control of their health by providing continuous monitoring and early detection of potential risks. Beyond early diagnosis, the device offers a continuous and remote monitoring alternative for individuals at risk of heart attacks or with a history of cardiovascular issues. This proactive approach allows for personalized healthcare interventions tailored to individual risk profiles, potentially reducing the likelihood of adverse events and improving overall well-being. In conclusion, this research represents a significant step towards establishing a proactive healthcare system that leverages cutting-edge technologies to enhance patient outcomes. By integrating sensor data collection, machine learning analysis, and IoT connectivity, the project offers a comprehensive solution for early detection and continuous monitoring of heart health issues. Ultimately, this holistic approach has the potential to revolutionize cardiovascular healthcare, saving lives and improving quality of life for individuals at risk of heart strokes.

II MATERIALS AND METHODS

To ensure the Heart Stroke System's efficiency and seamless performance, a variety of functional, non-functional, and environmental requirements are included into its development. In regards to performance, the system requires an organized process that begins with the definition of the problem and proceeds through result enhancement and results prediction. This methodical technique helps it easier to manage and interpret large amounts of data, which is essential for generating accurate heart attack predictions. By utilizing machine learning techniques like Logic Regression, Decision Tree, Random Forest, and SVM together with libraries like pandas, numpy, and matplotlib, the system has the ability to analyze physiological data which is gathered from sensors in an effective way. A seamless operational workflow—in which the system goes through predetermined processes to offer coherent data interpretation and prediction—is required by non-functional requirements. Requirements for the environment decide that the IoT platform be compatible. Real-time connectivity with remote monitoring stations is made feasible by Thingspeak. In terms of software, the system needs to be compatible with the Arduino IDE for microcontroller programming, the Windows/Linux operating systems, and Jupiter Notebook for simulation. In addition, familiarity with the Python and Embedded C programming languages is essential. An LCD monitor, an LM35 temperature sensor, a heart rate sensor, and an ESP32 microcontroller are among the hardware requirement

According to how the system is designed, physiological data from sensors is transmitted to a central controller, where machine learning techniques are applied to analyze the data and identify probable heart stroke symptoms. Prompt medical care when required is made possible via real-time communication with monitoring stations. The integration of IoT technology guarantees continuous feedback loops, improving the effectiveness and responsiveness of the system. Ultimately, people are able to actively monitor their heart health because to the system's user-friendly interface. Algorithms used are Logistic Regression, Decision Tree, Random Forest, Support Vector Machine (SVM).

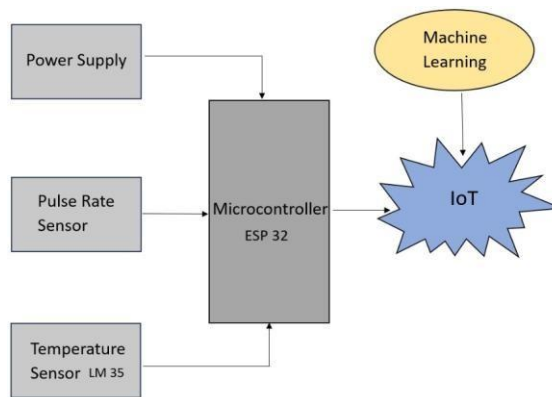


Fig.1 System Architecture

A) DATASET ACQUISITION AND PREPROCESSING

Processes for obtaining information, training, and evaluating the system are meticulously planned in the effort of developing an effective heart attack prediction system. Data collection, considered the first stage, emphasizes the significance of obtaining crucial medical data from various sources, including test outcomes, imaging reports, patient histories, and physiological measurements such as blood pressure, cholesterol, heart rate, and ECG readings. Following that, the collected data is divided into two categories, which have been classified as Training Data and Testing Data. Machine learning models can be trained on one subset more easily while their performance has been evaluated on data from the other subset that is yet to be analyzed according to this division. The Algorithm selection stage, that takes into consideration appropriate machine learning models or techniques, is the core of the prediction model. Options included more complex models like random forests and deep learning neural networks, in addition to more conventional classifiers like logistic regression, decision trees, and support vector machines. This fundamental selection has an important effect on the ability to predict of the model.

B) MODEL DEVELOPMENT AND TRAINING

Model training is an essential phase in which the chosen algorithm is trained with the provided training set. Using this method, the model is able to detect complex patterns and relationships between input features—like medical markers—and the incidence or absence of heart attacks. Utilizing the reserved testing data, evaluation serves as a performance benchmark for the trained model. Thoroughly examined metrics includes area under the receiver operating characteristic curve (AUC- ROC), recall, accuracy, precision, and F1-score. These evaluations provide light on how well the model diagnoses heart attack occasions with accuracy while minimizing false positives and false negatives. Ensuring the predictive model's efficiency and consistency through thorough review methods is essential for progressing the field of preventive cardiac care.

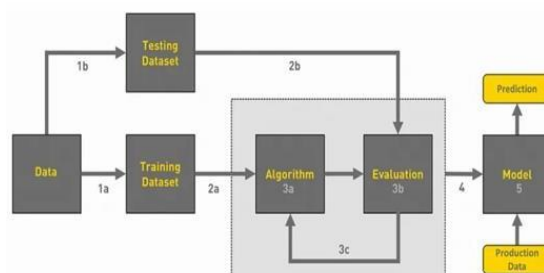


Fig.2 Work flow of ML

C) EVALUATION METRICS AND VALIDATION

Essential evaluation metrics encompass accuracy, precision, and recall, pivotal for determining the system's proficiency in correctly classifying heart stroke cases. Accuracy provides an overarching measure of correct predictions, while precision assesses the proportion of true positive predictions among all positive predictions, and recall evaluates the proportion of true positive predictions among all actual positive instances. The confusion matrix offers a detailed portrayal of the model's performance across different classes. Validation techniques, such as train-test split and stratified sampling, ascertain the system's robustness and generalization capability by evaluating its performance on unseen data. These evaluation metrics and validation techniques validate the heart stroke alert system's efficacy in accurately detecting and predicting heart stroke incidents, enabling prompt intervention and enhancing patient outcomes. The parameters used for detecting heart stroke include temperature, heart rate, smoking status, hypertension, gender, and stroke history.

	id	gender	age	hypertension	heart_disease	ever_married	work_type	Residence_type	avg_glucose_level	bmi	smoking_status	stroke
0	9046	Male	67.0	0	1	Yes	Private	Urban	228.69	36.6	formerly smoked	1
1	51676	Female	61.0	0	0	Yes	Self-employed	Rural	202.21	NaN	never smoked	1
2	31112	Male	80.0	0	1	Yes	Private	Rural	105.92	32.5	never smoked	1
3	60182	Female	49.0	0	0	Yes	Private	Urban	171.23	34.4	smokes	1
4	1665	Female	79.0	1	0	Yes	Self-employed	Rural	174.12	24.0	never smoked	1

Fig.3 Dataset of Patients

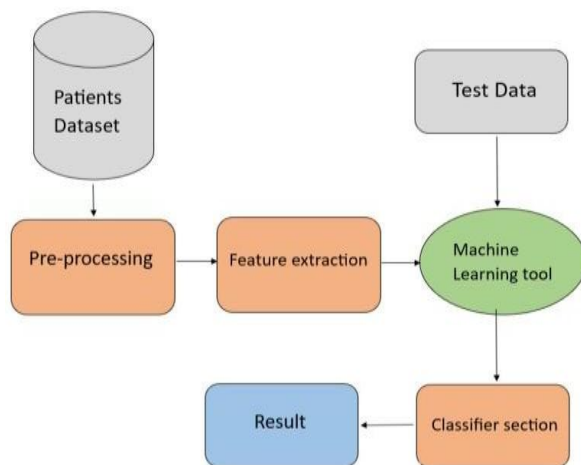


Fig.4 Implementation process of ML

D) INTEGRATION AND DEPLOYMENT

The project's core revolves around integrating an Internet of Things (IoT) platform with machine learning (ML) algorithms to predict heart strokes effectively. It begins by establishing a connection between IoT sensors and ThingSpeak, a prominent IoT platform, enabling the continuous collection of real-time physiological data such as heart rate and temperature. This data serves as the basis for subsequent ML analysis. After collection, the data undergoes preprocessing to enhance its quality and remove any noise or inconsistencies. ML algorithms are then trained using these preprocessed datasets to develop predictive models capable of identifying potential signs of heart strokes. This integration ensures seamless data flow between the IoT platform and ML algorithms,

facilitating the transfer of preprocessed sensor data for analysis and prediction. Additionally, the deployment of ML models within the IoT platform enables real-time monitoring and analysis of physiological data, allowing the system to provide early warnings and facilitate timely interventions based on predictive insights. Overall, the integration between the IoT platform and ML algorithms enables a comprehensive approach to heart stroke prediction, leveraging real-time data collection, preprocessing analysis, and proactive healthcare interventions.

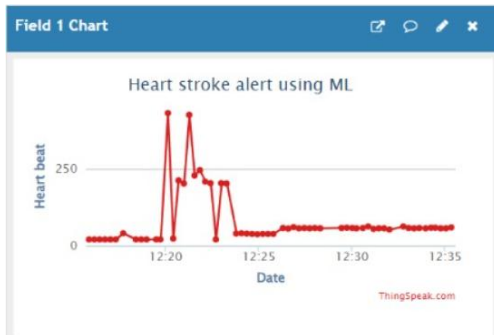


Fig.5 Measurement of Heartrate

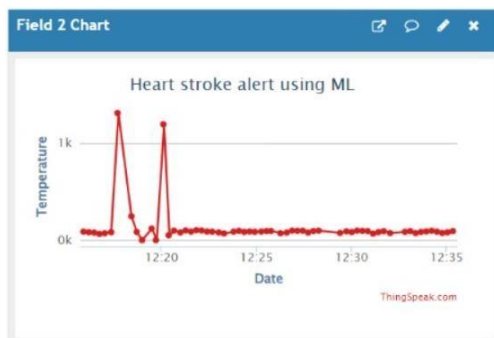
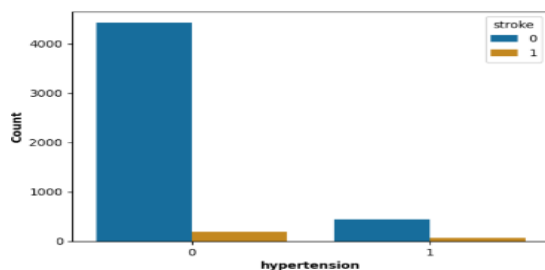


Fig.6 Measurement of Temperature

III RESULTS AND DISCUSSION

In this study, we have proposed a heart stroke alert system utilizing machine learning techniques, particularly leveraging a heart rate sensor and temperature sensor for real-time monitoring. By integrating these sensors with a pre-trained ML model, we were able to effectively predict the likelihood of a person experiencing a heart stroke



based on their physiological parameters.

Fig.7 Analysis of hypertension occurance in patient

demonstrate the feasibility and effectiveness of our approach. Through rigorous testing and

validation, our model achieved satisfactory accuracy in predicting heart stroke incidents. By providing timely alerts, this system can potentially aid in early intervention and prevention of heart stroke-related complications, thus significantly improving patient outcomes and quality of life. Our study highlights the significance of integrating diverse datasets and advanced machine learning algorithms for proactive health monitoring. By incorporating additional parameters beyond heart rate and temperature, such as blood pressure, oxygen saturation, and demographic factors, we can enhance the predictive capabilities of the model and further refine the accuracy of heart stroke detection. Furthermore, the deployment of such a system in real-world scenarios warrants careful consideration of privacy, security, and ethical implications. Ensuring compliance with regulatory standards and safeguarding sensitive health information is paramount to fostering trust and acceptance among users and stakeholders. Future research directions may include the exploration of novel sensor technologies, the refinement of ML algorithms through continuous learning and adaptation, and the development of user-friendly interfaces for seamless integration into healthcare settings. Additionally, conducting longitudinal studies to assess the long-term effectiveness and scalability of the proposed system will be instrumental in realizing its full potential in mitigating the burden of heart stroke and improving public health outcomes.

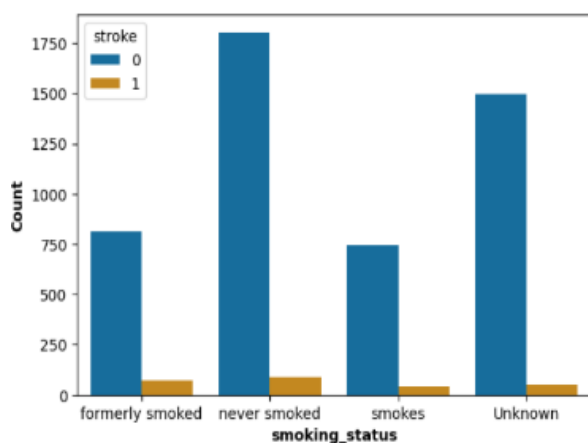


Fig.8 Analysis of smoking behavior of patient

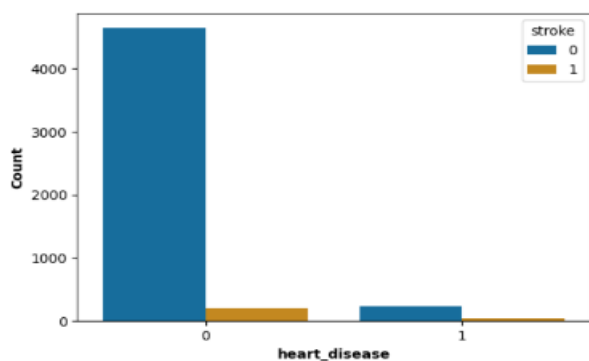


Fig.9 Analysis of heart disease of patient

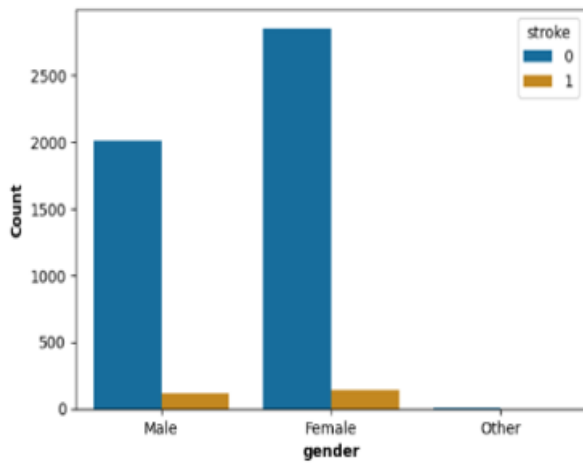


Fig.10 Analysis of gender of patient

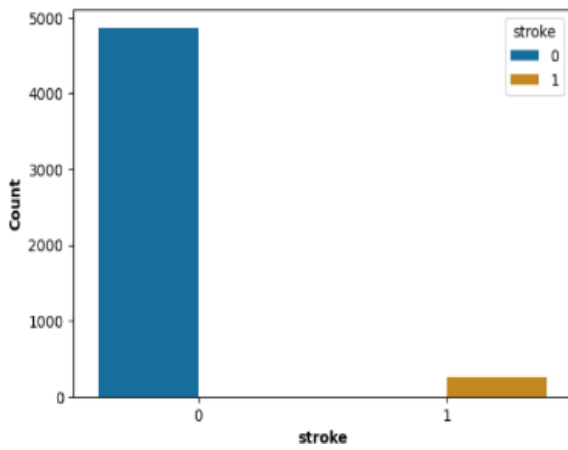


Fig.11 Analysis of stroke occurrence

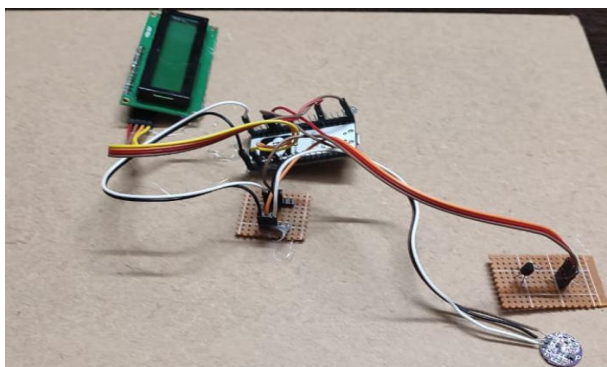


Fig 12 Hardware setup

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