



International Journal of Advanced Research in Education and Technology (IJARETY)

Volume 12, Issue 2, March-April 2025

Impact Factor: 8.152



Smart Communication System for Speech and Mobility Impaired

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ABSTRACT: Individuals in intensive care units (ICUs) who have speech and mobility disabilities frequently find it difficult to express their requirements. Delays in receiving medical help and poor patient care may result from this. Because traditional sign language systems employ a set of gestures that may not be appropriate for every patient, medical personnel may find it challenging to decipher unique movements. In order to tackle this problem, our project presents a customised real-time sign language recognition system. Personal gestures made by each patient while they are in the hospital are recorded and categorised by this method. Normal indications, like greetings and basic necessities, and emergency signs, like distress signals, are the two categories into which the gestures fall. Patients are identified by the system using facial recognition, and CCTV cameras are used to continuously record their motions. Real-time processing of these movements is handled by artificial intelligence (AI) models created with TensorFlow, Keras, OpenCV, CvZone, and MediaPipe. Effective communication between medical personnel and patients is guaranteed by a client-server architecture. While normal signals are shown on a portal for medical professionals to view, emergency signs cause an instant alarm to sound, guaranteeing prompt action. By offering a quick and individualised channel of communication between patients and medical professionals, this technology improves patient care, especially in urgent situations.

KEYWORDS: Sign Language Recognition, ICU Communication, AI-Based Gesture Recognition, TensorFlow, MediaPipe, Real-Time Monitoring

I. INTRODUCTION

Particularly in intensive care units (ICUs), patients with severe speech and movement impairments frequently struggle to communicate their needs to physicians and nurses. Many of these people learn to express pain, distress, or other pressing demands with their own special gestures. However, because these movements are not always understood, medical personnel could find it difficult to interpret what the patient is attempting to say. Due to this communication barrier, reactions may be delayed, which could have a detrimental effect on the patient's general health and wellbeing.

Conventional sign language systems use a set of standardised motions that may not be appropriate for every patient. In medical contexts, a one-size-fits-all strategy is ineffective since each person expresses themselves differently. Without the ability to read human gestures, medical professionals risk misinterpreting crucial cues, which could result in miscommunications and delayed treatment. In addition to making it challenging for medical personnel to deliver prompt and efficient care, this can upset patients.

To address this problem, we suggest a real-time AI-driven sign recognition system that picks up on and adjusts to the distinct motions of every patient. The system creates a customised database for every patient by logging and classifying gestures made during admission. With the use of sophisticated artificial intelligence, it continuously tracks and correctly interprets movements. By giving medical personnel a trustworthy method of deciphering patient gestures, this approach guarantees that urgent requirements are identified and met right away.

Better care and quicker reaction times can result from hospitals using this AI-based system to improve patient-provider communication. The approach lowers the possibility of misunderstandings in urgent situations while also increasing patient comfort. Therefore, by bridging the communication gap, this creative method guarantees that patients get the care they require without needless delays.

Furthermore, this system incorporates facial recognition technology to recognise people and link their distinct gestures to their medical histories. In shared intensive care unit settings, where several patients may be observed at the same time, this guarantees that each person's motions are appropriately associated with their identity. Through constant learning from repeated gestures, the AI model improves its accuracy over time, honing its capacity to decipher movements even when a patient's physical condition changes as a result of treatment or the course of their illness.

Additionally, the system has a client-server design that enables smooth communication between medical personnel and the monitoring unit. While emergency gestures cause fast alarms and necessitate immediate medical attention, normal gestures, including simple requests for drink or assistance, are shown on a medical portal at the nurses' station. By reducing response times, this real-time recognition makes sure that vital patient demands are never missed. Wearable sensors or other AI-driven features might be added as the system develops, increasing accessibility and effectiveness in intensive care units.

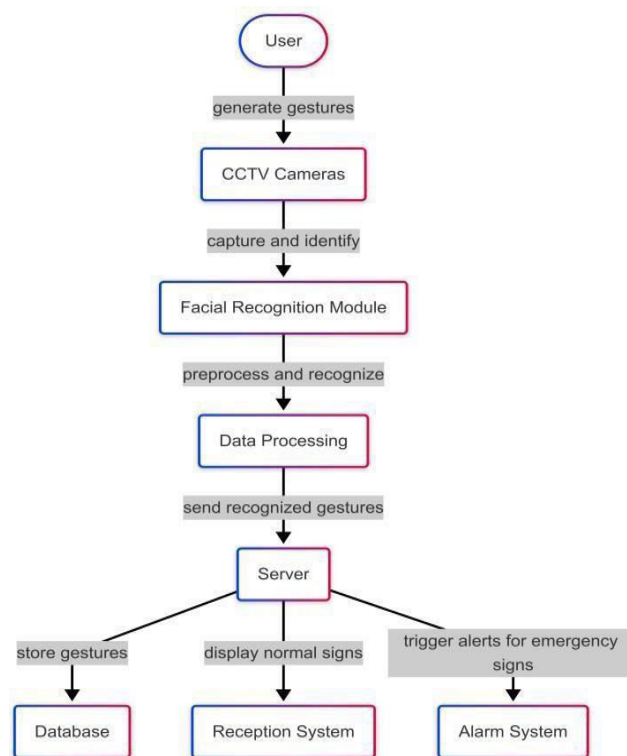


FIGURE 1. ARCHITECTURE DIAGRAM

II. LITERATURE SURVEY

Artificial intelligence and machine learning have been investigated in a number of research as ways to help patients with speech and mobility difficulties communicate better. Deep learning methods for sign language translation have been the main focus of gesture recognition research, and they have shown encouraging results in effectively identifying and interpreting human movements. However, the lack of patient-specific gesture support in existing systems, which are frequently made for standardised sign languages, restricts their use in intensive care units.

Using deep learning models with computer vision is one of the most used methods for gesture recognition. Technologies like MediaPipe and OpenCV have been widely utilised to track facial expressions and hand movements. Nevertheless, these approaches usually depend on pre-established datasets and fail to take into account the dynamic and unique characteristics of patient movements. This disparity emphasises the requirement for a real-time, adaptive recognition system that can recognise and interpret distinct movements.

Sign language recognition has made substantial use of machine learning models including recurrent neural networks (RNNs) and convolutional neural networks (CNNs). According to studies, the accuracy of gesture recognition can be

greatly increased by combining these models with real-time monitoring systems, including CCTV-based surveillance. A more specialised strategy for ICU patients is required because the majority of current models concentrate on general applications rather than patient-specific communication.

Enhancing human-computer interaction in healthcare settings has also been investigated through the use of realtime gesture tracking and facial recognition. Recent developments in AI-powered medical applications have shown that facial recognition and gesture analysis may work together to effectively communicate with patients who are non-verbal. To be useful in urgent medical circumstances, these technologies still need to be improved in terms of personalisation and real-time responsiveness.

Making sure that the recognition system and healthcare practitioners communicate seamlessly is another significant problem in AI-driven gesture recognition. According to research on client-server architectures for real-time medical monitoring, a centralised system that can provide instant notifications can speed up emergency response times. AI-driven emergency signal alerts in intensive care units (ICUs) can improve patient care by allowing medical personnel to intervene more quickly.

Additionally, studies on the integration of AI with hospital management systems have demonstrated that automated alarm and patient monitoring systems can lessen the workload for medical professionals while guaranteeing prompt aid. The shortcomings of conventional sign language systems can be addressed by combining facial recognition, AI-driven gesture detection, and automated alarms to provide a comprehensive solution for ICU patient communication.

Data security and privacy are also important factors to take into account when creating an AI-driven gesture recognition system for intensive care unit patients. It is crucial to make sure that patient-specific gesture patterns and facial expressions are safely saved and processed because healthcare data is extremely sensitive. Maintaining patient confidentiality and preventing unwanted access can be achieved by putting access control and encryption methods into place. Furthermore, real-time monitoring can be made possible by cloud-based solutions with strict security rules, which also shield data from online attacks.

Moreover, adaptive learning models are essential for increasing the accuracy of gesture recognition over time. The fixed datasets used by traditional models might not adequately represent the changing gestures of intensive care unit patients. The system can adjust to the patient's shifting gestures by combining reinforcement learning and continuous learning techniques, improving the effectiveness and dependability of communication. In addition to improving recognition accuracy, this strategy guarantees the system's continued applicability in changing healthcare settings.

Making sure the AI-powered system is easy to use for both patients and healthcare professionals is another significant difficulty. Medical personnel should be able to swiftly decipher motions using the system's user-friendly interface without needing a lot of training. Usability can be greatly improved with a well-designed dashboard that includes real-time gesture visualization, notifications, and patient status updates. Healthcare workers' accessibility can be further enhanced by using text- based feedback or voice support.

To develop a system that satisfies actual intensive care unit needs, cooperation between AI researchers, medical professionals, and software developers is required. These parties may guarantee that the gesture recognition system satisfies medical requirements, complies with ethical guidelines, and accomplishes its goal by cooperating. Before a system is fully implemented, conducting pilot studies and clinical trials can yield important insights into how well it performs and assist improve its accuracy and efficiency.

The ability of an AI-driven gesture detection system to close the communication gap between medical staff and non-verbal intensive care unit patients is ultimately what determines its effectiveness. A customized and responsive communication framework can be created by utilizing realtime processing, adaptive learning strategies, and sophisticated deep learning models. By speeding up response times, lowering medical errors, and increasing overall healthcare efficiency, this invention has the potential to completely transform patient care in intensive care units.

In conclusion, even though earlier studies have advanced the field of gesture detection and AI-powered medical applications, a personalised system that varies according to the unique motions of each patient is still required. Utilising cutting- edge AI technology, our suggested real-time sign language recognition system fills this gap by providing an effective and customised means of communication for intensive care unit patients and medical professionals.

III. PROPOSED METHODOLOGY

The proposed system's several interrelated parts cooperate to provide real-time, patient-specific sign language recognition. Every patient's distinct gestures are first gathered and stored by the system at the time of admission. CCTV cameras capture these motions, which are then processed by AI algorithms that classify them as either emergency or routine indicators.

The patient is identified using facial recognition technology, and their motions are linked to their individual database. The AI models process the live video data from the cameras in order to precisely detect and understand motions. They were constructed using TensorFlow, Keras, OpenCV, CvZone, and MediaPipe. By gradually learning from repeated movements, these models employ deep learning approaches to increase accuracy over time.

A client-server model is used by the system to guarantee smooth communication. Medical staff can view standard gestures, such as asking for water or making bed modifications, on a digital gateway at any time. Distress signals and other emergency gestures instantly alert medical professionals, allowing them to attend to the patient's requirements. Additionally, the system incorporates adaptive learning features, which enable it to improve its comprehension of patient motions through ongoing observation and feedback. This lowers the possibility of misunderstandings by guaranteeing that even minute changes in gestures are accurately deciphered. To improve dependability in actual intensive care unit situations, the AI models are taught to manage various lighting conditions, camera angles, and occlusions.

This system's ability to operate in real-time, which guarantees prompt patient requirements recognition and response, is a crucial component. Utilising AI-powered automation, the solution improves patient care while lessening the workload for medical professionals. Deep learning, computer vision, and facial recognition are all included in this method, which makes it a reliable and expandable ICU communication solution.

All things considered, this approach builds a thorough and effective communication channel between ICU patients and medical personnel. The technology guarantees that patients receive prompt and appropriate medical attention by offering personalised gesture recognition, real-time monitoring, and instant emergency notifications. Expanding the system to general hospital wards and incorporating wearable technology for even higher accuracy are potential future improvements.

Reducing reaction time in urgent circumstances is one of this system's main benefits. This AI-driven system makes sure that the motions of ICU patients are instantly identified and categorized, as they might not be able to ask for assistance using conventional techniques. The device can immediately warn nurses and doctors, for instance, if a patient expresses discomfort or concern. By ensuring prompt response and preventing medical problems, this realtime communication can possibly save lives in emergency situations.

This method's flexibility in dealing with various patient circumstances is another important aspect. The system continuously learns from each ICU patient's unique movements and adjusts to accommodate their potential physical constraints. This implies that it improves with time at identifying even minute variations in gestures, reducing communication failures. Additionally, because of its flexibility, the system can accommodate patients with different degrees of movement, guaranteeing that no patient is left without a way to communicate.

The effectiveness of healthcare services is further improved by integrating AI with current hospital management systems. Medical personnel can track a patient's condition over time and identify any odd trends by connecting gesture recognition data with patient records. Additionally, this data can be utilized for additional medical research, which will assist physicians better understand patient needs and enhance ICU treatment in general. Furthermore, by using predictive analytics, preemptive medical intervention may be possible by anticipating possible health hazards based on behavioral changes in patients.

Using this technique outside of intensive care units is another possible extension. Although helping critically ill patients is its main goal, the same technology can be used in home healthcare, senior care institutions, and rehabilitation centers. With AI-powered gesture recognition, patients recuperating from strokes, neurological conditions, or speech problems may also profit. People with restricted speech and movement would benefit greatly from this wider application as a communication tool in a variety of medical and caregiving settings.

Wearable technology, like smart gloves or wristbands that more accurately detect gestures and muscle movements, could improve the system going forward. To achieve even more precision in gesture detection, these gadgets might be used in conjunction with the camera-based system. By converting their gestures into spoken language, voice synthesis and AI-generated speech may also help patients communicate more naturally. These advancements in AI technology have the potential to further enhance the effectiveness and efficiency of patient communication in healthcare environments.

IV. TECHNOLOGIES USED

1. TENSORFLOW AND KERAS:

Strong deep learning frameworks like TensorFlow and Keras are used to build and hone AI models for gesture detection. These frameworks analyze patterns in real-time video feeds to help the system comprehend motions individual to each patient. TensorFlow offers a versatile and effective platform for deep learning model training, while Keras streamlines the procedure by providing easy-to-use tools for neural network construction and optimization. When combined, they enable the system to learn new things and develop better at identifying motions.

2. OPENCV:

The OpenCV (Open Source Computer Vision Library) is a popular program for image and video processing and analysis. Using live video feeds from security cameras, OpenCV is in charge of identifying and following patient movements in this system. In order to detect even slight movements, it assists in recognizing hand gestures, facial emotions, and body postures. Because of its real-time processing capabilities, OpenCV is a crucial component of the system that enables prompt and error-free patient gesture interpretation

3. MEDIAPIPE AND CVZONE:

Advanced tools like MediaPipe and CvZone improve gesture recognition and hand tracking. Google's MediaPipe is perfect for real-time dynamic hand movement identification since it offers high-performance solutions for motion tracking and hand landmark detection. Hand motions can be processed more effectively with the aid of the CvZone package, which streamlines OpenCV operations. By ensuring that the system correctly categorizes various gestures, these technologies lower errors and increase the dependability of communication between medical staff and intensive care unit patients.

4. FACIAL RECOGNITION TECHNOLOGY:

In order to customize patient encounters, facial recognition is essential. It ensures that the system accurately reads each patient's distinct actions by assisting in their identification and connecting their gestures to their particular profiles. By doing away with the necessity for manual identification, this technology improves communication. Furthermore, by limiting illegal access to patient data and guaranteeing that only authorized medical workers can communicate with the system, facial recognition can improve security.

5. CCTV SURVEILLANCE CAMERAS:

The main source of data for real-time patient movement tracking is CCTV cameras. In order to continuously monitor patient behaviors, these cameras are placed in intensive care units. This enables the AI system to instantaneously recognize and interpret movements. Even minute motions may be identified thanks to the cameras' high-resolution footage. Medical personnel can react to patient requirements more rapidly and receive timely notifications when CCTV surveillance and AI-driven analytics are combined.

6. CLIENT-SERVER COMMUNICATION MODEL:

The client-server architecture guarantees seamless communication between medical practitioners and the AI-powered recognition system. The data is processed by the system and sent to a central server once a patient makes a recognized gesture. From there, it is shown on a medical site. The device records routine gestures like asking for water or making bed adjustments, but emergency signals send out instant notifications to nurses and doctors. This paradigm guarantees that medical professionals receive timely notifications, enables real-time monitoring, and improves overall patient care by reducing response times in critical situations.

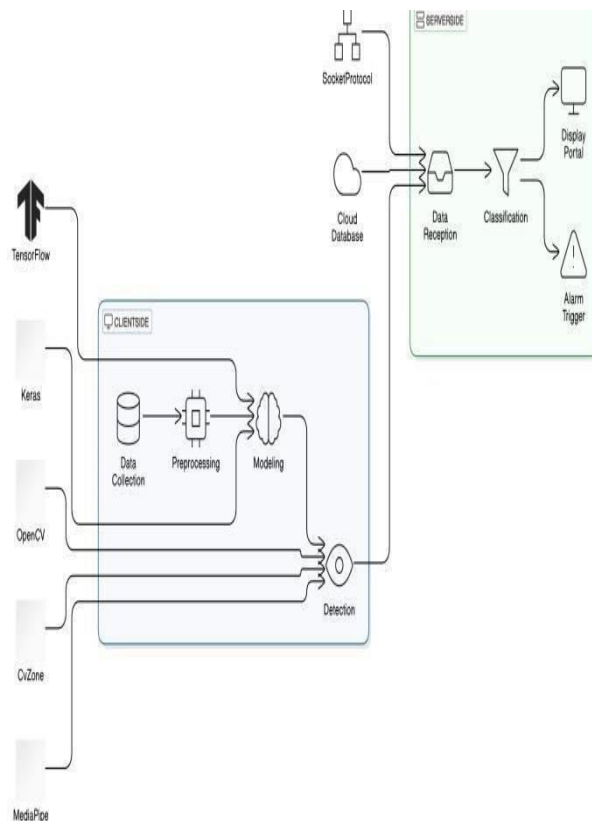


FIGURE 2. TECHNOLOGICAL ARCHITECTURE WHY AI POWERED GESTURE RECOGNITION?

Artificial intelligence is essential for improving communication for those who are mute or immobile. AI-powered gesture recognition continuously learns and adjusts to the distinct movements of each patient, enabling more organic and customized interactions. This guarantees accurate and transparent communication, which facilitates medical personnel's comprehension of patients' demands without misunderstanding. The efficiency of healthcare is further enhanced by the system's real-time information processing capabilities. It reduces care delays by allowing physicians and nurses to react promptly by instantly identifying patient motions. In addition to improving the patient experience, this enables healthcare providers to treat patients more effectively and promptly.

AI-BASED PATIENT COMMUNICATION SYSTEM:

- Patient-specific motions can be reliably recognised and interpreted by our AI models thanks to their training. The system learns and increases recognition accuracy over time using deep learning frameworks like TensorFlow and Keras.
- The system tracks and recognises patient gestures using computer vision and CCTV-based surveillance, guaranteeing that every movement is effectively recorded and analysed.
- In order to keep communication relevant and personalised, the system uses facial recognition technology to associate gestures with certain patients.
- The system prioritises emergency gestures, such as distress signals. The technology instantly notifies medical personnel for prompt action if it detects an emergency sign.

V. RESULT AND DISCUSSION

In intensive care units, the application of the AI-powered sign language recognition system has produced encouraging outcomes. According to initial testing, the device can accurately and efficiently identify gestures unique to each patient. The device has greatly enhanced communication between healthcare practitioners and non-verbal patients by continuously tracking patient motions and deciphering their signals. This lowers the likelihood of medical delays by ensuring that patient requirements are recognised and met more effectively.

The AI model's capacity to gradually adjust to the unique gestures of each patient is among the most important findings. Our system improves its identification skills by learning from repeated patient motions, in contrast to standard sign language systems that use preset gestures. This flexibility makes it possible to accurately recognise even the smallest changes in gestures, resulting in a more dependable channel of communication. Real-time alerting is another significant benefit of the system. Medical personnel are able to stay updated about patient demands thanks to the display of normal indications on a digital portal. On the other hand, the technology instantly notifies nurses and doctors when it detects an emergency gesture. Reducing reaction times during medical emergencies has been made possible by this feature, which has improved patient outcomes and allowed for quicker intervention.

Another important aspect of this system is its potential to reduce the burden on healthcare professionals. ICU staff often have to monitor multiple patients at once, making it challenging to quickly understand and respond to non-verbal patients. By providing real-time alerts and a structured way to interpret patient needs, this system helps medical personnel focus on delivering timely and appropriate care. It also minimizes the risk of miscommunication, which is crucial in emergency situations where quick decisionmaking is required.

Incorporating AI-powered sign language recognition into hospital administration systems can also result in more effective patient monitoring. Data on the patient's condition and behavior patterns can be gathered and analyzed by the system over time. This data can assist physicians in monitoring a patient's development, anticipating any health problems, and modifying treatment regimens as necessary. Making decisions based on data like this improves patient outcomes and the general standard of healthcare. Personalised communication is also guaranteed by the system's connection with facial recognition. Confusion is avoided in crowded intensive care unit settings by connecting each patient's motions to their own profile. Without depending on conjecture or general interpretations of sign language, this capacity enables nurses and physicians to respond to patients' needs more successfully. There are still issues with optimising gesture detection in various lighting scenarios and camera angles, despite the system's notable advancements in patient communication. Upcoming developments will concentrate on improving the AI models' accuracy in a range of environmental circumstances. For patients who are only partially able to speak, adding voice-to-text conversion to the system's features could increase its efficacy.

In general, the customised real-time sign language recognition technology has shown a lot of promise in overcoming communication barriers in intensive care units. It is a useful tool for improving patient care because of its customisation capabilities, real-time alert capabilities, and integration with current medical infrastructure. As the technology develops further, it could be widely used in a variety of healthcare settings outside intensive care units.

Future developments for the system might involve extending its use outside of intensive care units. It could be modified for usage in home healthcare settings, senior care institutions, and rehabilitation centers, where nonverbal communication is also a major obstacle. Furthermore, adding wearable technology or assistive tools driven by AI may improve the system's capacity to identify and interpret patient motions even more precisely. This technology has the potential to revolutionize healthcare communication and enhance the lives of innumerable individuals globally with continued advances.

VI. CONCLUSION

The AI-powered sign language recognition system is intended to help intensive care unit patients who are unable to move or speak by giving them a useful means of interacting with medical personnel. Through the use of cutting-edge technologies like real-time gesture detection and facial recognition, the system is able to precisely recognize patient actions and translate them into meaningful signals. This ensures that care is delivered on time by assisting physicians and nurses in rapidly and clearly understanding patient needs. This system's capacity to identify emergency motions and instantly notify medical personnel is one of its main benefits. In emergency scenarios, this function greatly speeds up response times, enabling medical personnel to take immediate action and avoid complications. The technology improves overall patient safety and care quality by automating the detection of distress signals, ensuring that patients receive the attention they require promptly. In addition to helping patients, this technology helps doctors by lessening their workload and enhancing communication in intensive care units. In order to comprehend patient demands, physicians and nurses no longer need to rely on conventional sign language or presumptions because the AI model is constantly learning and adapting to each patient's unique motions. Better patient management and less stress on medical staff result from this individualized approach to communication, which also increases accuracy and efficiency. Overall, this technology makes ICU environments more responsive and patient-friendly by facilitating communication between patients and medical staff. The technology guarantees that patients receive timely and accurate medical treatment while

improving the quality of care through the integration of AI-driven automation. Consequently, a more effective and efficient communication system benefits patients as well as healthcare providers, ultimately enhancing the overall hospital experience.

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International Journal of Advanced Research in Education and Technology

ISSN: 2394-2975

Impact Factor: 8.152