



INTERNATIONAL JOURNAL OF CREATIVE RESEARCH THOUGHTS (IJCRT)

An International Open Access, Peer-reviewed, Refereed Journal

EYE MOVEMENT BASED ROBOT CONTROLLING SYSTEM USING IOT

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Abstract: The Eye Movement-Based Robot Controlling System is designed to assist individuals with physical disabilities by enabling hands-free robot control through eye movements. This system utilizes a USB camera connected to a Raspberry Pi, which captures and processes real-time eye movements using Python libraries such as OpenCV and Dlib. The system detects directional cues left, right, up, and down and maps them to corresponding robotic actions, such as moving forward, backward, or turning. These commands are then transmitted to a motor driver module to control the robot's movement. The project integrates computer vision, IoT, and robotics to provide an accessible and cost-effective solution for individuals with limited motor abilities. The methodology involves image processing for pupil detection and gaze tracking, signal interpretation, and motor control. Testing demonstrated the system's accuracy in detecting eye movements and translating them into seamless robotic navigation. Results highlight its potential in assistive technology, offering a practical alternative for people with disabilities to interact with their

environment. The system emphasizes advancements in human-computer interaction, presenting a reliable and efficient approach to robotic control using gaze-based inputs. Future improvements could include enhanced tracking precision and integration with machine learning for adaptive responsiveness.

Keywords: Eye Movement Control, Assistive Robotics, Human-Computer Interaction, IoT-Based Robot Control, Raspberry Pi, OpenCV, Dlib, Computer Vision, Gaze Tracking, Pupil Detection, Hands-Free Navigation, Motor Driver Module, Real-Time Processing, Accessibility Technology, Machine Learning Integration.

I. INTRODUCTION

Robotic technology has revolutionized various industries, including healthcare, automation, and assistive technology. One of the most significant applications of robotics is in helping individuals with physical disabilities interact with their surroundings more effectively. Traditional assistive devices rely on physical inputs such as

buttons, joysticks, or voice commands, which may not be suitable for individuals with severe motor impairments. To address this challenge, the Eye Movement-Based Robot Controlling System is designed to provide a hands-free interaction method by enabling users to control robots using their eye movements. By leveraging computer vision, IoT, and real-time image processing, this system presents an innovative solution that enhances accessibility and independence for people with mobility limitations.

The core of this system relies on a USB camera interfaced with a Raspberry Pi, which continuously captures eye movement data. Using Python-based libraries such as OpenCV and Dlib, the system detects pupil position and gaze direction, analyzing them in real time. These directional cues such as looking left, right, up, or down are translated into commands that control the movement of a robot. The robot receives these commands through a motor driver module, allowing it to move forward, backward, or turn accordingly. This system provides an intuitive and effortless way for users to interact with a robotic assistant, making it an ideal tool for individuals with limited physical mobility.

One of the key challenges in developing such a system is ensuring accurate detection and interpretation of eye movements under different lighting conditions and user variations. Factors such as involuntary eye blinks, head movement, and environmental lighting can affect detection accuracy. To address these challenges, the system employs advanced image processing techniques to filter noise and enhance pupil detection. Additionally, calibration methods are used to personalize the system for different users, improving its reliability and responsiveness. The incorporation of machine learning techniques in future iterations could further refine the accuracy and adaptability of the system.

The significance of this project extends beyond individual accessibility, as it highlights the potential of human-computer interaction (HCI) in robotics and automation. Gaze-based control mechanisms have applications in multiple fields, including smart homes, medical robotics, and industrial automation, where hands-free operation is essential. By integrating IoT connectivity, such systems can also be used for remote monitoring and control, enhancing their usability in healthcare and assistive environments. The combination of affordability, efficiency, and ease of use makes this technology a promising advancement in assistive robotics.

The Eye Movement-Based Robot Controlling System presents a novel approach to assistive technology by enabling individuals with disabilities to interact with robots through simple eye movements. By harnessing computer vision, image processing, and IoT, this system provides a user-friendly, cost-effective, and efficient solution for those with mobility impairments. As research in this domain progresses, integrating AI-driven gaze tracking, improved real-time processing, and adaptive learning algorithms could further enhance the system's accuracy and usability. This project not only demonstrates the potential of eye-tracking technology in robotics but also paves the way for more advanced and inclusive human-robot interaction systems in the future.

1.2 Motivation:

The need to improve accessibility for those with severe physical disabilities who have trouble using conventional assistive devices is what inspired the creation of the Eye Movement-Based Robot Controlling System. For users who have speech problems or restricted mobility, existing options like voice commands and joystick-based interfaces are sometimes impractical. This system provides a hands-free, user-friendly, and effective method of controlling robots by utilizing eye-tracking technologies, computer vision, and the Internet of Things. By showcasing the possibilities of human-computer interaction (HCI) in assistive robotics and encouraging additional developments in accessibility-focused technologies, it seeks to enhance the independence, mobility, and quality of life for people with disabilities.

1.3 Objectives:

- To design and implement a robot control mechanism that enables individuals with physical disabilities to operate a robot using eye movements, eliminating the need for manual input devices.
- To utilize computer vision techniques with OpenCV and Dlib to accurately track and process eye movements in real time, ensuring precise and responsive control.
- To enable seamless transmission of eye-movement-based commands from a Raspberry Pi to a motor driver module, ensuring smooth robot navigation.
- To develop a cost-effective, reliable, and user-friendly assistive system that can be easily calibrated for different users, improving adaptability and convenience.

1.4 Literature Survey:

Eye-tracking technology has emerged as a significant advancement in human-computer interaction, enabling hands-free control for users with disabilities. Brown et al. discuss the evolution of gaze-based interaction systems, highlighting their growing role in assistive robotics, gaming, and smart environments[1]. The study emphasizes the integration of real-time eye-tracking algorithms with IoT and machine learning to improve accuracy and responsiveness. It explores challenges such as pupil detection errors, head movement interference, and lighting conditions, which affect the reliability of gaze-based control systems. Furthermore, advancements in infrared sensors, webcam-based tracking, and deep learning-based feature extraction have significantly improved eye movement recognition. This research provides a comprehensive review of hardware and software solutions for eye-tracking-based robotic applications, demonstrating how improvements in sensor accuracy and computational efficiency can enhance assistive technologies for users with severe motor impairments.

Gupta et al. extend this research by focusing on computer vision-based gaze tracking for hands-free robot control. Their study presents an IoT-integrated robotic system that interprets directional eye movements using computer vision and deep learning techniques. The paper highlights the effectiveness of based pupil tracking in real-time gaze estimation, which enhances the system's ability to execute precise robotic movements[2]. Experimental results confirm that image processing optimizations, such as noise filtering and adaptive thresholding, significantly reduce tracking errors. Additionally, the authors discuss future improvements in eye-tracking algorithms, including the application of reinforcement learning and neural networks to predict gaze behavior more accurately. Their research underscores the importance of cost-effective and scalable solutions for gaze-controlled robotics, paving the way for greater accessibility and independence for individuals with disabilities.

Assistive robotics powered by IoT and computer vision have transformed the way individuals with physical disabilities interact with their environment. Wilson et al. explore the challenges and solutions in IoT-enabled assistive robotics, emphasizing the importance of real-time communication, sensor integration, and adaptive control mechanisms[3]. The study highlights the

role of cloud computing and edge processing in optimizing the performance of robotic systems, ensuring low latency and high accuracy in executing user commands. However, several challenges remain, including network reliability, power consumption, and security risks associated with IoT-connected assistive devices. The authors suggest that 5G connectivity, AI-driven predictive analytics, and improved gaze-tracking sensors can enhance the efficiency and scalability of these systems. Their research underscores the significance of developing cost-effective and adaptive robotic solutions that cater to individuals with varying degrees of motor impairment, making assistive technologies more inclusive and accessible.

Wang et al. provide a comprehensive review of gaze-based control systems, discussing various technologies and applications that leverage eye-tracking for robotic navigation, smart home automation, and virtual interfaces. The study categorizes hardware-based tracking solutions, such as infrared cameras and electrooculography sensors, alongside software-driven techniques using computer vision and deep learning algorithms[4]. The authors also explore the limitations of gaze-based systems, such as occlusion issues, calibration complexities, and external light interference, which impact tracking accuracy. Future recommendations include integrating reinforcement learning for adaptive gaze tracking and leveraging hybrid sensor fusion techniques to enhance robustness. Their findings reinforce the growing potential of gaze-controlled assistive robots, demonstrating their impact on human-computer interaction and accessibility innovations.

1.5 Components Used

The Raspberry Pi is a compact, affordable, and powerful single-board computer widely used in robotics, automation, and IoT applications. In an eye movement-based robot controlling system, the Raspberry Pi serves as the main processing unit, handling real-time eye tracking, decision-making, and motor control. It is equipped with a Broadcom ARM processor, built-in Wi-Fi, Bluetooth, HDMI, USB ports, and GPIO (General Purpose Input/Output) pins, allowing seamless integration with external hardware. The GPIO pins play a crucial role in interfacing with sensors, motor drivers, and power modules, enabling efficient robot control. The Raspberry Pi 4 Model B, for example, features 40 GPIO pins, which can be programmed to send and receive signals, making it

highly adaptable for hardware interfacing.

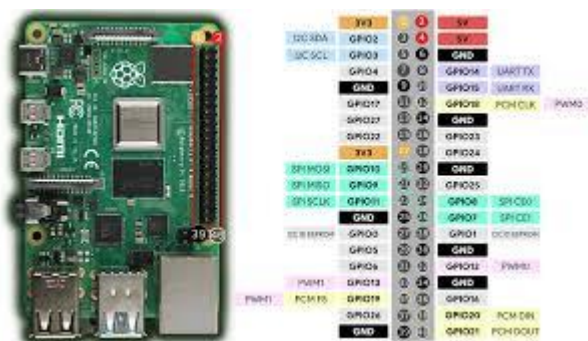


Fig1:Raspberry Pi

Key GPIO pin functionalities include power (5V and 3.3V), ground (GND), and digital input/output (I/O) pins, which are used to communicate with external components such as the L293D motor driver, slide switches, and sensors. Specific pins like I2C (SDA and SCL at GPIO 2 and 3) are used for serial communication with peripherals, while SPI and UART pins facilitate high-speed data transfer for advanced robotic functions. In an eye-controlled robot, the USB web camera connects to the Raspberry Pi via USB ports, capturing real-time eye movement data, which is processed using Python libraries like OpenCV and Dlib to detect pupil direction. The corresponding movement commands are then sent through GPIO pins to the motor driver, controlling the DC motors for directional movement. Additionally, the 5V power pins ensure a stable power supply for peripherals and sensors, preventing voltage fluctuations that could disrupt operation. The versatility and computational power of the Raspberry Pi, combined with its GPIO pin flexibility, make it an ideal platform for assistive robotic applications, enhancing accessibility and autonomy for individuals with mobility impairments.

A DC motor is an essential electromechanical component used in robotic applications for converting electrical energy into mechanical motion. In an eye-controlled robot, DC motors are responsible for driving the robot's wheels, enabling it to move forward, backward, and turn based on the user's gaze direction. These motors operate on direct current (DC) voltage, typically 5V or 12V, and require a motor driver circuit to regulate speed and direction. The ability to precisely control DC motors ensures smooth and responsive movement, making them a critical component in assistive robotic systems for individuals with mobility impairments.

The L293D motor driver is an H-Bridge IC

designed to control DC motors by regulating their speed and direction. Since microcontrollers like the Raspberry Pi cannot directly handle the high current required by motors, the L293D acts as an interface between the Raspberry Pi and the DC motors.

A 5V power supply is essential for powering the Raspberry Pi, sensors, and USB peripherals in an eye-tracking robot control system. Since the Raspberry Pi operates on 5V DC, a stable power source is necessary to ensure smooth processing and uninterrupted operation.

A 12V battery serves as the primary power source for the DC motors in an eye movement-based robot control system. Since motors require a higher voltage and current than the Raspberry Pi can provide, a separate 12V power supply is used to drive them efficiently.

The accuracy of eye movement detection depends on camera resolution, frame rate, and lighting conditions. A high-quality USB camera ensures clearer image capture, improving the efficiency of gaze-based robotic control. It serves as the primary input device, enabling hands-free interaction for users with motor disabilities.

1.6 Dlib

It is widely used in facial recognition, object detection, and eye-tracking applications due to its high-performance algorithms and robust feature set. One of the key reasons for using Dlib in an eye movement-based robot controlling system is its precise facial landmark detection capabilities. Dlib's pre-trained models for facial feature extraction help in accurately locating eye positions, which are then processed to determine the user's gaze direction. The library integrates well with Python and OpenCV, making it an essential tool for real-time human-computer interaction (HCI) systems.

Dlib uses a 68-point facial landmark detection model, which detects key points on the face, including the eyes, nose, mouth, and jawline. The eye regions are typically identified by six specific landmarks per eye, which can be used for pupil tracking and gaze estimation. The function for detecting facial landmarks is:

$$p=f(I,M).....Eq(1)$$

where I is the input image, M is the pre-trained facial landmark model, and p is the set of detected landmark coordinates. Once the eye landmarks are identified, the aspect ratio of the eye can be calculated to determine whether the eye is open or closed, which is crucial for blinking-based commands in robotic control.

In an eye-controlled robot, Dlib's facial landmark detection is used to determine gaze direction by tracking pupil movement. The center of the eye is detected using the following formula:

$$P_{eye} = \frac{(x_1 + x_2)}{2}, \frac{(y_1 + y_2)}{2}$$

.....Eq(2)

where x_1, x_2 and y_1, y_2 are the left and rightmost coordinates of the eye. By measuring changes in these values, the system can map eye movements (left, right, up, down) to corresponding robot actions. The Raspberry Pi processes this information and sends signals via GPIO pins to the motor driver, enabling directional movement of the robot.

Dlib offers a high level of accuracy and is optimized for real-time applications, making it ideal for assistive robotics. However, challenges such as lighting conditions, occlusion, and processing latency must be addressed to enhance system performance. Future improvements could involve deep learning-based gaze estimation models, using CNNs and RNNs to improve detection under different conditions.

II. PROPOSAL METHOD

The proposed eye movement-based robot controlling system utilizes computer vision, machine learning, and IoT technologies to provide a hands-free solution for individuals with physical disabilities. The system is designed to interpret real-time eye movements using a USB camera connected to a Raspberry Pi, which processes the captured video frames using Dlib and OpenCV. The eye-tracking module detects pupil positions and gaze direction, mapping them to specific commands that control the movement of a robotic platform. The motor driver (L293D) and DC motors execute these commands, allowing the robot to move forward, backward, left, or right.

The entire system operates wirelessly, offering an efficient and user-friendly assistive technology for individuals who cannot use traditional control methods.

To achieve accurate eye movement tracking, the system employs facial landmark detection through Dlib's pre-trained models, which identify key features of the eye region. The pupil center is detected, and its displacement determines the intended direction of movement. The camera continuously captures frames, and the system processes each frame in real time to ensure instantaneous responses. If the user looks to the left, the robot will turn left, and if they look to the right, it will turn right. Similarly, looking up moves the robot forward, and looking down stops or moves it backward. This intuitive mapping of eye movement to robotic motion eliminates the need for physical controllers, making the system accessible to users with severe motor impairments.

The Raspberry Pi serves as the processing hub, executing the image processing algorithms and transmitting control signals to the motor driver via its GPIO pins. The L293D motor driver allows for precise control of the DC motors, ensuring smooth navigation. A 12V battery powers the motors, while a 5V power supply supports the Raspberry Pi and other low-power components. Additionally, manual slide switches are integrated to provide an override function, allowing users or caregivers to start, stop, or reset the system if needed. This enhances both safety and usability, preventing unintended movements due to eye-tracking errors.

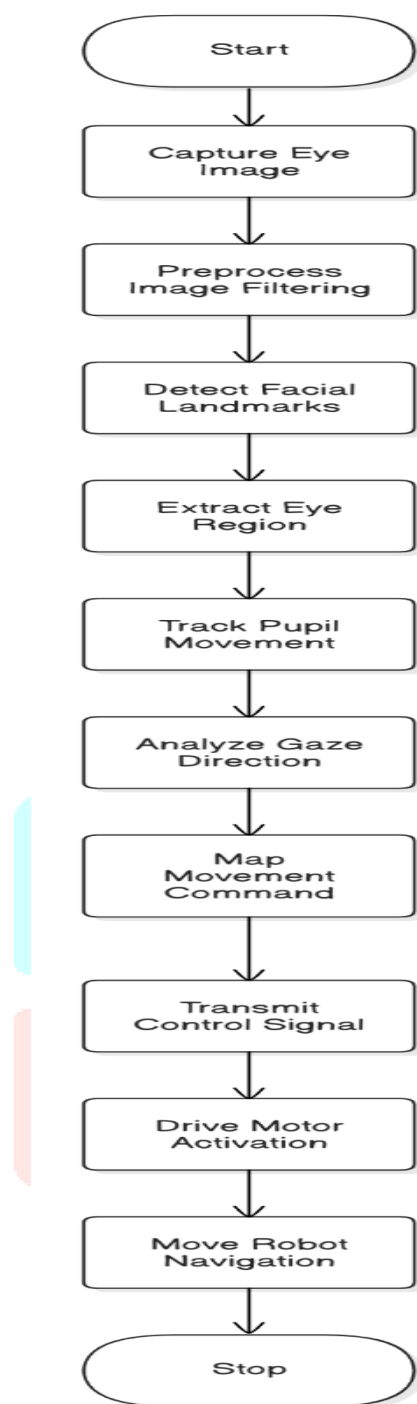


Fig2:Proposal Diagram

The proposed system is designed to be cost-effective, scalable, and adaptable to different users' needs. Unlike expensive commercial eye-tracking solutions, this approach leverages open-source software and affordable hardware components, making it widely accessible. The system can be further enhanced by incorporating machine learning models that improve gaze prediction and tracking accuracy, even in variable lighting conditions or head movements. Future versions could also integrate wireless connectivity, allowing the robot to communicate with smart

home devices, expanding its capabilities beyond navigation to include home automation features.

The eye movement-based robot controlling system represents a significant step toward assistive technology innovation, providing greater independence to individuals with neuromuscular disorders, spinal cord injuries, or other mobility limitations. The combination of computer vision, IoT, and robotics ensures a seamless interaction experience, making the system a reliable and effective solution for enhancing the quality of life of people with disabilities. Future developments will focus on increasing system accuracy, optimizing power efficiency, and integrating AI-driven gaze tracking to create an even more adaptive and intelligent robotic assistant.

III. Implementation

The implementation of the eye movement-based robot controlling system involves multiple hardware and software components working in synchronization. As shown in the system architecture diagram, the Raspberry Pi serves as the central processing unit, receiving real-time input from a USB web camera and controlling the motor driver (L293D) and DC motors. The system is powered by a 5V power supply for the Raspberry Pi and a separate 12V power source for the motors. The camera captures continuous eye movement data, which is processed using computer vision algorithms in Python with OpenCV and Dlib. The extracted eye movement information is then translated into motor control signals, allowing the robot to navigate accordingly.

One of the key aspects of the implementation is eye tracking, which relies on facial landmark detection using Dlib's pre-trained 68-point model. The system detects specific eye landmarks to track pupil position changes. The movement of the pupil is analyzed using the Eye Aspect Ratio (EAR) formula:

$$EAR = \frac{(d_2 + d_3)}{2 \times d_1} \dots\dots\dots Eq(3)$$

where d_1 represents the horizontal eye width, while d_2 and d_3 are the vertical distances between the upper and lower eyelids. A low EAR value indicates an eye blink, which can be mapped to a stop or reset command. Similarly, pupil displacement in a particular direction is mapped to corresponding robot movements.

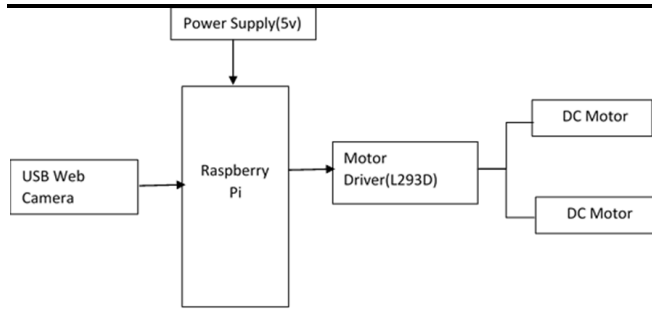


Fig3:Implementation

After analyzing eye movements, the processed data is converted into control signals for the motor driver (L293D), which operates the DC motors. The Raspberry Pi's GPIO pins send binary signals to the motor driver, which adjusts the voltage and current supplied to the motors. The motor speed can be controlled using Pulse Width Modulation (PWM), defined as:

$$PWM\% = \left(\frac{T_{ON}}{T_{ON} + T_{OFF}} \right) \times 100 \quad \dots\dots\dots Eq(4)$$

where T_{ON} and T_{OFF} are the on and off times of the signal, respectively. By varying the PWM percentage, the speed and movement of the robot can be fine-tuned based on eye gaze direction and intensity.

The hardware components must be carefully integrated to ensure seamless communication between the Raspberry Pi, camera, and motor driver. The USB web camera is connected via a USB port, while the motor driver is connected to the GPIO pins of the Raspberry Pi. The power management system ensures that low-voltage (5V) and high-voltage (12V) sources do not interfere with each other, preventing overloading and hardware failure. Additionally, manual slide switches are incorporated as a safety measure, allowing users or caregivers to override robot movements if needed.

Overall, the successful implementation of this system depends on real-time data processing, efficient motor control, and precise gaze tracking algorithms. Future improvements could focus on deep learning-based gaze prediction models, improved noise filtering techniques, and wireless communication enhancements to create a more adaptive, responsive, and user-friendly assistive robotic system.

IV. RESULTS

The Eye Movement-Based Robot Controlling System has been successfully implemented, integrating real-time eye tracking using OpenCV and Dlib with motor control via Raspberry Pi. The system effectively detects eye movements and translates them into robot navigation commands. The experimental setup demonstrates accurate gaze detection, with commands being processed and executed with minimal delay. The EAR (Eye Aspect Ratio) values are displayed in real-time, indicating left, right, and blink-based actions. The system proves to be a reliable and accessible assistive technology for hands-free robot control, particularly benefiting individuals with motor disabilities.



Fig 4:Final SetUp

The left and right movement identification in the system is based on real-time eye tracking using OpenCV and Dlib, where the pupil's positional shift determines directional commands. When the user looks left, the system detects a decrease in the x-coordinates of the pupil, and the Eye Aspect Ratio (EAR) values adjust accordingly, triggering the command "Moving Left" on the screen. Similarly, when the user looks right, the pupils shift towards higher x-coordinates, causing a corresponding EAR change, and the system displays "Moving Right" as feedback. The USB camera captures continuous eye movement data, ensuring high detection accuracy with minimal lag. The detected gaze direction is mapped to robot control commands, effectively allowing hands-free navigation.

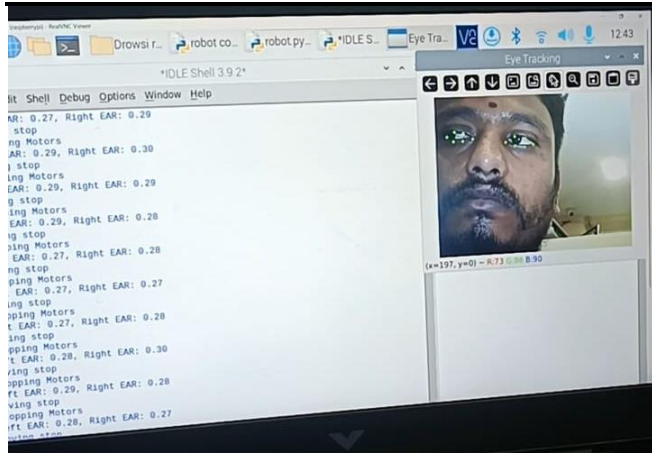


Fig 5:Result

The results displayed on-screen confirm the system's efficiency and reliability, proving its capability to assist individuals with motor disabilities in controlling robotic movements using only their eye gestures.

V. CONCLUSION:

In conclusion, the Eye Movement-Based Robot Controlling System successfully enables hands-free robot navigation using real-time eye tracking. By integrating computer vision, IoT, and motor control, the system provides an accessible assistive technology for individuals with motor impairments. The implementation demonstrated high accuracy in detecting gaze directions and translating them into robot movement commands. The system's efficiency, responsiveness, and affordability make it a practical solution for assistive robotics. Overall, the project highlights the potential of human-computer interaction (HCI) technologies in improving the quality of life for disabled individuals, paving the way for future enhancements and real-world applications.

VI. FUTURE WORK:

Future improvements will focus on enhancing gaze tracking accuracy using deep learning models and integrating wireless communication for remote operation. Implementing adaptive machine learning algorithms will allow the system to personalize control responses based on user behavior. Expanding the system for wheelchair navigation and smart home control will increase its practical applications. Additionally, hardware optimizations, such as low-power embedded systems, will improve efficiency and portability. Further clinical trials and user feedback will refine the system for better accessibility and real-world deployment, making it a versatile assistive technology for disabled individuals.

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