





SMART WHEELCHAIR WITH BRAIN-COMPUTER INTERFACE (BCI) CONTROL

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Abstract

This paper outlines the design and implementation of an innovative smart wheelchair system aimed at advancing mobility solutions for those with severe physical disabilities. The system integrates a Brain-Computer Interface (BCI) based on electroencephalography (EEG) with supplementary control modalities including voice commands, eye gaze tracking, a traditional joystick, and head tilt detection via a gyroscope. A Raspberry Pi serves as the central processing unit, orchestrating data acquisition from various sensors, processing control signals, and executing motor commands for navigation. Safety is paramount, with ultrasonic sensors implemented for real-time obstacle detection and avoidance. Furthermore, Internet of Things (IoT) connectivity via platforms like Firebase or ThingSpeak enables remote monitoring and emergency alert functionalities. The system employs Artificial Intelligence (AI) and machine learning algorithms for robust processing of EEG and vision signals, striving for a reliable, adaptable, and user-centric assistive device. This research demonstrates





the potential of a multi-modal control approach to significantly improve the quality of life and independence of individuals with limited motor capabilities.

1 Introduction:

Being able to move freely is a basic part of human independence and quality of life. For those who face serious mobility limitations resulting from conditions such as conditions like spinal cord injuries or ALS (Lou Gehrig's disease), or cerebral palsy, traditional methods of mobility, including manual wheelchairs or caregiver assistance, can significantly limit their autonomy and participation in daily activities. The development of intelligent assistive devices, particularly smart wheelchairs, offers a promising avenue to address these limitations.

Existing smart wheelchair technologies often rely on single or limited control interfaces, which might not work well for everyone or all situations. For instance, those affected by progressive disorders of the nervous system may experience fluctuations in their motor abilities, necessitating alternative control mechanisms. We've designed this project to help solve this problem by proposing a novel smart wheelchair system that integrates multiple control modalities, providing users with a flexible and adaptable means of navigation and interaction with their environment.

The core of this system lies in the integration of an EEG-based Brain-Computer Interface (BCI), enabling direct control through brain activity. To enhance usability and provide redundancy, the system also incorporates voice commands for simple directional control and environmental interactions, eye gaze tracking for intuitive visual target selection, a conventional joystick for those with a bit of remaining hand function, and head tilt detection via a gyroscope for subtle directional inputs. The Raspberry Pi acts as the central hub, processing data from these diverse inputs and translating them into precise motor commands for the wheelchair.

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Safety is a critical consideration in the design. Ultrasonic sensors are integrated to provide real-time obstacle detection, enabling this setup to autonomously avoid collisions. What's more, the system's IoT connection lets users monitor it remotely of the user's location and wheelchair status, along with the capability to trigger emergency alerts in critical situations. The application of Artificial Intelligence and machine learning algorithms is crucial for the effective processing of complex EEG signals and visual data from the eye gaze tracking system, ensuring reliable and accurate control.

This paper details the design, development, and integration of these various components into a cohesive smart wheelchair system. It outlines the hardware and software architecture, the signal processing techniques employed, and the safety and IoT functionalities implemented. The potential impact of this multi-modal approach on enhancing the independence and daily comfort and independence for those with serious mobility limitations is also discussed.

2 Literature Review:

The field of smart wheelchairs has witnessed significant advancements in recent years, exploring various control interfaces and functionalities. This section provides an overview of relevant research in BCI-based wheelchair control, voice control, eye gaze tracking, and multi- modal integration.

2.1 Brain-Computer Interface (BCI for Wheelchair Control:

BCIs offer a direct communication pathway between the brain and external devices, bypassing traditional neuromuscular pathways. Electroencephalography (EEG) is a widely used non-invasive BCI modality for wheelchair control due to its relatively low cost and ease of use. Research in EEG-based wheelchair control has explored various paradigms, including motor imagery (imagining movements), event-related potentials (ERPs) like P300,





and steady- state visual evoked potentials (SSVEPs). While significant progress has been made, challenges remain in achieving robust and reliable control in real-world environments due to the inherent noise and variability in EEG signals.

2.2 Voice Control for Wheelchairs:

Voice recognition technology offers a hands-free control interface that can really make a difference for individuals facing limited upper body mobility. Existing voice-controlled wheelchairs typically allow users to issue simple commands for navigation (e.g., "forward," "left," "stop") and potentially for interacting with smart home devices. However, voice control can be susceptible to environmental noise and might not be the best fit for everyone, especially those with speech difficulties.

2.3 Eye Gaze Tracking for Wheelchair Control:

Eye gaze tracking technology allows users to control devices by simply looking at specific targets on a screen. In the context of smart wheelchairs, eye gaze allows users to choose directional commands or navigate virtual environments. Advancements in computer vision and eye-tracking algorithms have made this control modality increasingly accurate and reliable. However, factors such as user fatigue and the need for calibration can pose challenges.

2.4 Multi-Modal Control in Smart Wheelchairs:

Recognizing the limitations of single-interface control, researchers have explored the integration of multiple control modalities in smart wheelchairs. Combining different input here approaches come with a few benefits, such as increased robustness, adaptability to varying user abilities and environmental conditions, and enhanced user experience. For example, a system might allow a user to primarily navigate using BCI while utilizing voice





 $commands \ for \ specific \ actions \ or \ switching \ to \ joystick \ control \ if \ BCI \ performance \ degrades.$

2.5 Safety and IoT Integration in Smart Wheelchairs:

Safety is paramount while creating any assistive device. Smart wheelchairs often incorporate sensors like ultrasonic or infrared sensors that help detect obstacles and collision avoidance. Furthermore, integrating IoT connectivity enables features like remote monitoring of the user location and health status, as well including the option to send emergency alerts to caregivers or medical professionals.

This project builds upon the existing body of knowledge by proposing a comprehensive multi-modal smart wheelchair system that synergistically combines the strengths of EEG-based BCI, voice control, eye gaze tracking, joystick input, and head tilt detection, while also incorporating robust safety features and IoT connectivity.

3 System Architecture and Design:

The proposed smart wheelchair system comprises several key modules that work in concert to provide enhanced mobility and safety. The overall system architecture is as shown in Figure 1.

(Figure 1: System Architecture Diagram - This would visually represent the interconnected modules: EEG Sensor, Voice Recognition Module, Eye Tracker, Gyroscope, Joystick, Raspberry Pi (Central Processing Unit), Motor Control Unit, Wheelchair Motors, Ultrasonic Sensors, IoT Module (Firebase/ThingSpeak), Power Supply.)

3.1 Hardware Components:

• EEG Sensor: A non-invasive EEG headset is used to acquire brainwave information coming from the user's scalp. The specific type of headset will be chosen based on factors like how many channels are being used, signal quality, and ease of use.







- Voice Recognition Module: A microphone and a speech processing unit are used to capture and interpret voice commands from the user.
- Eye Tracker: A camera-based eye tracking system is employed to monitor the user's gaze direction on a display or the environment.
- Gyroscope: A miniature gyroscope sensor is mounted on the user's head to detect head tilt movements.
- Joystick: A standard wheelchair joystick is integrated to provide an alternative control method for people who still have a bit of hand function.
- Raspberry Pi: A Raspberry Pi single-board computer acts as the brain of the system. It collects data from all the sensors, signal processing, decision-making, and communication with the motor control unit and the IoT module.
- Motor Control Unit: It gets directions from the Raspberry Pi and controls the speed and direction of the wheelchair motors.
- Wheelchair Motors and Drive System: Standard DC motors and a suitable drive mechanism are used to propel the wheelchair.
- Ultrasonic Sensors: Multiple Ultrasonic sensors placed around the wheelchair help it detect obstacles nearby, making navigation safer.
- IoT Module: A Wi-Fi or cellular module is used to connect the Raspberry Pi to the internet for remote monitoring and emergency alerts via platforms like Firebase or ThingSpeak.
- Power Supply: A reliable power supply system is implemented to power all electronic components.

3.2 Software Components:

• EEG Signal Processing: Software algorithms are implemented on the Raspberry Pi to preprocess, filter, and classify EEG signals to extract relevant control commands based on the chosen BCI paradigm (e.g., motor imagery classification using machine learning).







- Voice Command Processing: Speech recognition software is used to listen and translate speech into commands the system can follow which are then interpreted to control wheelchair actions.
- Eye Gaze Tracking Software: Software processes the video feed from the eye tracker to determine the user's gaze direction and map it to specific control commands or targets.
- Gyroscope Data Processing: Algorithms are used to analyze the gyroscope data and translate head tilt movements into directional control signals.
- Sensor Fusion and Control Logic: A central control algorithm on the Raspberry Pi integrates the inputs from all active control modalities based on user preference or system-defined priorities. This module also incorporates obstacle avoidance logic based on data from the ultrasonic sensors.
- Motor Control Interface: Software on the Raspberry Pitranslates the processed control commands into signals that are sent to the motor control unit to drive the wheelchair motors.
- IoT Communication: Software libraries are used to establish communication with the chosen IoT platform (e.g., Firebase SDK or ThingSpeak API) for data logging, remote monitoring, and emergency alert functionalities.
- User Interface: A user interface (potentially a touchscreen display) may be integrated to allow users to select control modes, calibrate sensors, and view system status.

4 Implementation Details:

The implementation of the smart wheelchair system involves several key stages:

4.1 Hardware Integration:

The various hardware components are physically integrated onto a standard wheelchair frame. This includes mounting the sensors (EEG headset interface, microphone, eye tracker, gyroscope, ultrasonic sensors), the Raspberry Pi, the motor control unit, and ensuring proper





wiring and power distribution. Considerations for ergonomics, user comfort, and safety are paramount during this stage.

4.2 Software Development:

The software development involves implementing the tools that process and interpret signals for each control modality, the sensor fusion and control logic, the motor control interface, and the IoT communication protocols. Machine learning methods that analyze and classify signals from the brain and vision processing for eye gaze tracking are trained and deployed on the Raspberry Pi. The choice of programming languages (e.g., Python, C++) and software libraries (e.g., for signal processing, machine learning, computer vision, and IoT communication) is crucial for efficient development and performance.

4.3 Multi-Modal Control Integration:

What really stands out in this project is how well all the different ways of controlling it are combined. This involves developing a strategy for how the system handles simultaneous inputs from different sensors and how people can smoothly switch from one mode to another control modes. Potential approaches include user selection via a menu or automatic switching based on signal quality or user state.

4.4 Obstacle Detection and Avoidance:

The Raspberry Pi is linked to the ultrasonic sensors to pick up distance information, and algorithms are implemented to process the distance data. The control logic incorporates obstacle avoidance behaviours such as automatically slowing down or stopping when an obstacle is detected in the wheelchair's path. More advanced algorithms for path planning around obstacles may also be explored.





4.5 IoT Connectivity and Remote Monitoring:

The Raspberry Pi is configured to connect to the chosen IoT platform (e.g., Firebase or ThingSpeak). Data such as the wheelchair's location (obtained via GPS if integrated), battery level, and potentially sensor readings can be logged and visualized remotely. An emergency alert mechanism is implemented, allowing the user or the system to trigger notifications to caregivers or emergency services.

5 Artificial Intelligence and Machine Learning

Applications:

AI and machine learning play an important role in enabling robust and adaptive control:

- EEG Signal Processing: Machine learning algorithms, such as Support Vector Machines (SVM), Convolutional Neural Networks (CNNs), or Recurrent Neural Networks (RNNs), are employed to classify different ways the brain responds during specific user intentions (e.g., move forward, turn left, stop). Training these models requires collecting EEG data from individual users performing the desired mental tasks.
- Eye Gaze Tracking: Computer vision algorithms and potentially we use machine learning to help us to accurately figure out where the user is looking using the video feed This may involve calibration procedures to map eye movements to screen coordinates or environmental targets.
- Sensor Fusion: AI techniques could be explored to intelligently f use input from multiple sensors to respond more accurately, improving the reliability and accuracy of the overall control system. For example, a Kalman filter or other state estimation techniques can help to combine noisy sensor readings.
- Adaptive Control: Machine learning could be used to personalize the control system based on individual user characteristics and preferences. For instance, the sensitivity of the different control interfaces could be changes on its own as it learns from user performance.





6 Safety Considerations:

Safety is a paramount concern while creating and building this smart wheelchair. The following safety features are implemented:

- Obstacle Detection and Avoidance: Ultrasonic sensors provide real-time information about the environment, enabling the system to automatically avoid collisions.
- Emergency Stop Mechanism: A readily accessible physical emergency stop button is integrated into the system.
- Fail-Safe Control: In the event of a system malfunction or loss of communication with a primary control interface, the system may automatically switch to a more basic control mode (e.g., joystick) or come to a safe stop.
- Battery Monitoring: The system monitors the battery level and provides alerts to the user when recharging is necessary.
- Remote Monitoring and Emergency Alerts: IoT connectivity allows caregivers to remotely monitor the user's location and receive emergency alerts if the user requires assistance.

7 Future Improvements:

While the proposed system offers a significant advancement in smart wheelchair technology, several avenues for future improvement exist:

- Advanced AI Models: Exploring more sophisticated deep learning models for EEG and vision signal processing could lead to improved accuracy and robustness.
- Gesture Recognition: Integrating gesture recognition using cameras or other sensors could provide an additional intuitive control modality.
- Environmental Interaction: Expanding the system's capabilities to interact with smart home environments or other external devices through voice commands or other interfaces.
 - Personalized Control Adaptation: Implementing more advanced machine learning





techniques to continuously adapt the control system to the user's evolving needs and preferences.

- Haptic Feedback: Incorporating haptic feedback mechanisms could provide users with additional sensory information about their movement and the environment.
- Navigation in Complex Environments: Developing more sophisticated path planning algorithms to enable autonomous navigation in complex and dynamic environments.

8 Conclusion:

Creating and developing a multi-modal smart wheelchair integrated with a Brain-Computer Interface holds immense potential to revolutionize mobility solutions to support those with serious mobility impairments. By combining the intuitive control offered by BCI with the redundancy and adaptability of voice commands, eye gaze tracking, joystick input, and head tilt detection, this system aims to provide users with a more reliable, versatile, and user-friendly means of navigation and interaction with their surroundings. The integration of ultrasonic obstacle detection and IoT connectivity further enhances the safety and provides valuable remote monitoring capabilities. The application of AI and machine learning is crucial for processing complex sensor data and enabling adaptive control. Future advancements in these areas promise to further enhance the capabilities and usability of such smart assistive devices, ultimately contributing to greater independence and an improved enhancing everyday life for those living with mobility impairments.