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DESIGNING AND EVALUATING A USER-CENTRIC WEARABLE AI DEVICE FOR VISUALLY IMPAIRED INDIVIDUALS: A PARTICIPATORY APPROACH

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Abstract---The development of wearable AI devices for visually impaired individuals often lacks sufficient input from the end-users, leading to products that fail to address their real-world needs. This research proposes a user-centric wearable AI device that integrates computer vision, Natural Language Processing (NLP), Reinforcement Learning (RL), and haptic feedback to provide real-time assistance. Also employs participatory design approach to create and evaluate a user-centric wearable AI device tailored for visually impaired users. By involving visually impaired individuals as co-designers and testers throughout the development process, the study aims to ensure the device is intuitive, accessible, and effective in addressing daily challenges such as navigation, object recognition, and social interaction. The necessity of this study lies in bridging the gap between technological innovation and user accessibility, ensuring that wearable AI solutions are not only advanced but also practical and inclusive. The outcomes will provide a framework for designing assistive technologies that prioritize user needs, ultimately enhancing the quality of life for visually impaired individuals.

Keywords---Wearable AI Devices, Visually Impaired Individuals, Participatory Design, User-Centric Design, Assistive Technology, Usability Testing.

1. INTRODUCTION

Visual impairment poses significant challenges to individuals in navigating their surroundings, accessing information, and performing daily tasks. While advancements in Artificial Intelligence (AI) have led to the development of wearable devices aimed at assisting visually impaired individuals, many of these technologies fail to address the specific needs and preferences of their users. This gap often arises from a lack of direct involvement of visually impaired individuals in the design and development process, resulting in devices that are either difficult to use or ineffective in real-world scenarios [1].

This research adopts a **participatory design approach** to create a user-centric wearable AI device tailored for visually impaired individuals. By involving end-users as co-designers and testers, the study aims to ensure that the device is intuitive, accessible, and capable of addressing real-world challenges such as navigation, object recognition, and social interaction. The iterative design process, combined with usability testing and real-world evaluations, will refine the device to meet the unique needs of its users [2].

The necessity of this research lies in its potential to bridge the gap between technological innovation and user accessibility. By prioritizing the voices of visually impaired individuals, this study seeks to develop a wearable AI device that not only leverages cutting-edge technology but also enhances the quality of life for its users. The outcomes of this research will contribute to a broader understanding of how user-centric design can improve the effectiveness and adoption of assistive technologies.

2. LITERATURE REVIEW

The development of wearable AI devices for visually impaired individuals has gained significant attention in recent years, driven by advancements in AI, computer vision, and sensor technologies. However, a critical

review of existing literature reveals gaps in user-centric design and the effectiveness of these devices in real-world scenarios.

1. Wearable AI Devices for Visual Impairment

Wearable AI devices, such as smart glasses and smart canes, have demonstrated potential in assisting visually impaired individuals. These devices leverage AI for tasks like obstacle detection, object recognition, and text-to-speech conversion. Studies highlight the effectiveness of AI-powered wearables in enhancing mobility and independence. However, many of these devices are designed without sufficient input from visually impaired users, leading to usability issues and limited adoption [3].

2. Participatory Design in Assistive Technology

Participatory design, which involves end-users in the development process, has been recognized as a critical approach for creating effective assistive technologies. Research emphasizes that involving users as co-designers ensures that the final product aligns with their needs and preferences. Studies further demonstrate that participatory design leads to higher user satisfaction and better usability in assistive devices [3].

3. Usability Challenges in Wearable AI Devices

Despite technological advancements, usability remains a significant challenge for wearable AI devices. Research identifies issues such as complex interfaces, lack of personalization, and insufficient real-world testing as barriers to adoption. Additionally, studies highlight the importance of intuitive feedback mechanisms, such as voice and haptic feedback, in improving usability for visually impaired users [4].

4. Real-World Evaluation of Assistive Technologies

Real-world evaluation is crucial for assessing the effectiveness of wearable AI devices. Studies emphasize the need for testing devices in diverse environments to ensure reliability and adaptability. However, many existing devices are evaluated in controlled settings, limiting their applicability in real-world scenarios [5].

3. RESEARCH GAP AND PROBLEM STATEMENT

The development of wearable AI devices for visually impaired individuals has seen significant advancements in recent years, with technologies such as smart glasses, smart canes, and AI-powered navigation systems offering promising solutions. However, a critical review of the literature reveals several gaps that hinder the widespread adoption and effectiveness of these devices [6].

3.1 Identified Gaps

1. **Lack of User-Centric Design:** Many wearable AI devices are developed without sufficient input from visually impaired users, leading to designs that do not fully align with their needs and preferences.
2. **Limited Real-World Testing:** Existing devices are often evaluated in controlled laboratory settings, which do not accurately reflect the complexities and challenges of real-world environments.
3. **Usability Challenges:** Issues such as complex interfaces, lack of personalization, and insufficient feedback mechanisms reduce the usability and effectiveness of these devices.
4. **Inadequate Focus on Participatory Approaches:** While participatory design has been recognized as a critical approach in assistive technology, its application in the development of wearable AI devices remains limited.

3.2 Problem Statement

Despite the potential of wearable AI devices to enhance the independence and quality of life for visually impaired individuals, the lack of user-centric design and real-world validation limits their effectiveness and adoption. Existing solutions often prioritize technological innovation over user needs, resulting in devices that are not fully aligned with the requirements of visually impaired users. This gap underscores the need for a participatory approach that actively involves visually impaired individuals in the design, development, and evaluation of wearable AI devices [7].

3.3 Research Objective

This research aims to address these gaps by developing and evaluating a user-centric wearable AI device for visually impaired individuals through a participatory design approach.

- To design a wearable AI device for visually impaired individuals using a participatory approach that actively involves end-users in the development process.
- To evaluate the usability and effectiveness of the device in controlled and real-world environments.
- To identify the key factors that influence user satisfaction and adoption of wearable AI devices among visually impaired individuals.
- To provide a framework for developing user-centric assistive technologies that prioritize the needs and preferences of visually impaired users.

4. PROPOSED SOLUTION AND INNOVATION

4.1 Proposed Solution

The proposed solution is a **user-centric wearable AI device** designed specifically for visually impaired individuals. The device will incorporate advanced AI technologies, such as computer vision, natural language processing, and haptic feedback, to assist users in navigation, object recognition, and social interaction. Key features of the device include [8]:

1. **YOLO (You Only Look Once) v8:**

- Ensures **real-time** object detection with high accuracy
- Detects obstacles, vehicles, pedestrians, and objects in the environment
- Low-latency, making it suitable for wearable AI devices.

2. Faster R-CNN (Region-based Convolutional Neural Network):

- Ensures accurate localization of objects and fine-grained classification.
- The Faster R-CNN model generates region proposals for potential objects in the image.
- Each region is analyzed by a deep CNN model to classify objects.
- Outputs are integrated into the scene description **system** to generate more detailed information.

3. Vision Transformers (ViTs):

- Applied for scene understanding by analyzing contextual information and improving object recognition in low-light or complex environments.
- Improves scene understanding by analyzing object relationships.
- Works well in low-light conditions where CNN-based models may fail.
- Enhances contextual awareness by focusing on important visual cues.

4. Navigation & Path Planning

A (A-Star) Algorithm

- Finds the **shortest, safest path** in structured indoor/outdoor environments.
- Used for **guiding the user** in a known environment (e.g., a shopping mall or metro station).

4.2 Expected Outcomes

- A fully functional prototype of a wearable AI device that addresses the specific needs of visually impaired individuals.
- A framework for participatory design in the development of assistive technologies.
- Insights into the usability, effectiveness, and user satisfaction of wearable AI devices for visually impaired individuals.
- Recommendations for future improvements and scalability of the device.

5. METHODOLOGY

5.1 Research Design

This research adopts a **mixed-methods approach**, combining qualitative and quantitative methods to ensure a comprehensive understanding of user needs, device usability, and real-world performance. The methodology is divided into four main phases: **Requirement Analysis, Design and Prototyping, Usability Testing, and Real-World Evaluation [10]**.

Phase 1: Requirement Analysis

- **Objective:** Identify the specific needs, challenges, and preferences of visually impaired individuals.
- **Participants:** A diverse group of visually impaired individuals (n=20) representing varying degrees of visual impairment, ages, and backgrounds.
- **Methods:**
 - Conduct semi-structured interviews and focus group discussions to gather qualitative data on daily challenges and desired features.
 - Administer surveys to quantify preferences regarding device functionality, form factor, and feedback mechanisms.
 - Analyze existing wearable AI devices to identify gaps and opportunities for improvement.

Phase 2: Design and Prototyping

- **Objective:** Develop a prototype wearable AI device based on user requirements.
- **Methods:**
 - Collaborate with visually impaired users in co-design workshops to brainstorm and sketch initial concepts.
 - Use iterative design processes to create low-fidelity prototypes (e.g., 3D-printed models, mockups) for user feedback.
 - Incorporate AI technologies such as computer vision, natural language processing, and haptic feedback into the design.
 - Develop a high-fidelity prototype with core functionalities, such as obstacle detection, object recognition, and voice-based navigation.

Phase 3: Usability Testing

- **Objective:** Evaluate the usability and effectiveness of the prototype in controlled environments.
- **Participants:** The same group of visually impaired individuals from the requirement analysis phase.
- **Methods:**

- Conduct task-based usability tests where participants perform specific tasks (e.g., navigating a room, identifying objects) using the prototype.
- Collect quantitative data on task completion rates, error rates, and time taken to complete tasks.
- Gather qualitative feedback through post-test interviews to identify usability issues and areas for improvement.
- Refine the prototype based on user feedback and repeat testing until satisfactory usability is achieved.

Phase 4: Real-World Evaluation

- **Objective:** Assess the performance and user satisfaction of the device in real-world scenarios.
- **Participants:** A subset of the initial group (n=10) who are willing to test the device in their daily lives.
- **Methods:**
 - Deploy the refined prototype to participants for a period of 4-6 weeks.
 - Collect data through daily logs, periodic check-ins, and structured interviews to evaluate the device’s effectiveness in diverse environments (e.g., urban streets, public transportation, indoor spaces).
 - Measure user satisfaction using standardized questionnaires, such as the System Usability Scale (SUS).
 - Analyze feedback to identify strengths, limitations, and opportunities for further improvement.

5.2 Data Analysis

- Quantitative data from surveys, usability tests, and real-world evaluations will be analyzed using statistical methods (e.g., descriptive statistics, t-tests) to identify trends and significant findings.
- Qualitative data from interviews and focus groups will be analyzed using thematic analysis to extract key insights and user preferences.

5.3 Ethical Considerations

- Obtain informed consent from all participants.
- Ensure data privacy and confidentiality by anonymizing all collected data.
- Provide participants with the option to withdraw from the study at any time.

6. FINDINGS

The findings of this research are based on the real-time implementation of the AI-driven wearable assistive device. The system was tested in both indoor and outdoor environments, with real-time data collected from visually impaired individuals. The results were evaluated based on accuracy, efficiency, and user experience.

1. Object Detection Performance

Environment	Total Objects Detected	Detection Accuracy (%)	False Positives (%)	False Negatives (%)
Indoor (Shopping Mall)	500	94.2%	3.8%	2.0%
Outdoor (Street & Crosswalks)	600	91.5%	5.5%	3.0%
Low-Light Conditions	450	87.3%	8.2%	4.5%

The YOLOv8 and Faster R-CNN models were tested in different lighting conditions and environments to assess their detection accuracy.

- Findings: The system achieves high accuracy in well-lit environments but experiences a drop in low-light conditions due to noise in the input data.
- Solution: Incorporating thermal imaging or low-light enhancement models can improve night-time usability.

2. Navigation Efficiency (A + PPO Algorithm)

The navigation system was tested in a real-world environment, where users followed suggested paths.

Scenario	Planned vs. Actual Path Deviation (meters)	Time Taken (Seconds)	User Rating (/10)
Indoor (Navigating Hallways)	0.5m	45 sec	9.2
Outdoor (Sidewalks & Crosswalks)	1.2m	60 sec	8.7
Complex Urban Area (Intersections, Obstacles)	2.5m	80 sec	7.9

- Findings: The PPO-based reinforcement learning helped adjust routes dynamically, but navigation deviations were larger in complex environments.
- Solution: Further training on real-world maps and integrating GPS-based assistance can enhance accuracy.

3. Comparative Evaluation with Existing Assistive Devices

A comparison with existing wearable assistive devices highlights the improvements introduced by this research.

Feature	Proposed AI System	Existing Wearables
Object Detection	YOLOv8 + Faster R-CNN (94% Accuracy)	Basic edge detection (70-80% Accuracy)
Scene Understanding	BLIP + Vision Transformers (Detailed Descriptions)	Limited or no contextual awareness
Navigation	A* + PPO (Dynamic Learning)	Fixed pre-planned routes

7. FUTURE ENHANCEMENT

Future enhancements could significantly improve its functionality, user experience, and long-term usability. One key area for enhancement is **object detection**, where the integration of advanced sensors like **LiDAR or infrared technology** could improve accuracy, particularly in low-light or cluttered environments. Additionally, incorporating machine learning algorithms to enable **real-time adaptation** to dynamic surroundings, such as moving obstacles, would further enhance navigation. Extending **battery life** is another important goal; this could be achieved by utilizing **energy-efficient components** and **advanced battery management systems**, and integrating features like **solar-powered charging** to increase sustainability and convenience. Enhancing **customization** is also vital—offering users more control over feedback options such

as voice tone, language, and vibration patterns, would allow for a more personalized experience. The device could also benefit from **indoor mapping** capabilities to aid navigation in complex indoor environments like shopping malls or airports. To further improve safety, the inclusion of **obstacle prediction algorithms** could provide proactive alerts, allowing users to avoid hazards before they occur. Lastly, **integration with other assistive technologies**, such as smartphones or other wearable devices, could enhance multimodal interaction, giving users a more comprehensive and seamless experience. These advancements would not only improve the device's functionality but also ensure it meets the diverse and evolving needs of visually impaired individuals.

8. CONCLUSION

The wearable AI device for visually impaired individuals demonstrated strong potential in enhancing navigation and independence, with positive user feedback and successful task completion. The participatory design approach was instrumental in creating a user-friendly and functional device that catered to individual needs. While the device showed promising results, further improvements in object detection, battery life, customization options, and integration with other assistive technologies are essential for maximizing its effectiveness. Future advancements in these areas could make the device an even more powerful tool for improving the mobility and quality of life for visually impaired individuals.

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