

# Object Detection for the Visually Impaired

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**Abstract:** *This paper presents the design and development of a mobile application, built using Flutter, that leverages object detection to enhance the lives of visually impaired individuals. The application addresses a crucial challenge faced by this community – the lack of real-time information about their surroundings. We propose a solution that utilizes pre-trained machine learning models, potentially through TensorFlow Lite for on-device processing, to identify objects within the user's field of view captured by the smartphone camera. The application goes beyond simple object recognition. Detected objects are translated into natural language descriptions through text-to-speech functionality, providing crucial auditory cues about the environment. This real-time information stream empowers users to navigate their surroundings with greater confidence and independence. Accessibility is a core principle of this project. The user interface will be designed with compatibility for screen readers, ensuring seamless interaction for users who rely on assistive technologies. Haptic feedback mechanisms will be incorporated to provide non-visual cues and enhance the user experience. The ultimate goal of this project is to create a user-friendly and informative application that empowers visually impaired people to gain greater independence in their daily lives. The application has the potential to improve spatial awareness, foster a sense of security, and promote overall inclusion within society.*

**Keywords:** Object detection, Voice Feedback

## I. INTRODUCTION

Vision impairment is a significant global challenge, affecting millions of people worldwide. According to the World Health Organization (WHO), over 2.2 billion people have a vision impairment, with at least 1 billion having moderate or severe distance vision loss. This can significantly impact an individual's ability to navigate their surroundings, perform daily tasks, and interact with the environment. In recent years, advancements in mobile technology and artificial intelligence (AI) have opened new avenues for assistive technologies that can empower visually impaired individuals. One particularly promising area is object detection, which utilizes machine learning models to identify and classify objects within an image or video stream. Several research efforts have explored the potential of object detection for visually impaired people. In their 2022 study, Jha et al. demonstrated the effectiveness of a deep learning-based object detection system for real-time obstacle detection, improving navigation safety for visually impaired users. Similarly, a 2021 study by Mancilla-Roa et al. proposed a convolutional neural network (CNN) based approach for object recognition in indoor environments, aiding visually impaired users with way finding tasks. This paper builds upon this existing research by presenting the design and development of a mobile application, specifically created using the Flutter framework, that leverages object detection to enhance the lives of visually impaired individuals. Our application goes beyond simple object recognition, aiming to provide users with a real-time stream of auditory information about their surroundings through text-to-speech functionality. This information can significantly improve spatial awareness and foster a sense of independence for visually impaired users. The remainder of this paper is structured as follows. Section 2 details the system design and implementation of the mobile application. Section 3 discusses the user interface design considerations, particularly focusing on accessibility features. Section 4 outlines the potential impact and future directions for this project. Finally, Section 5 concludes the paper by summarizing the key contributions of this research.

## II. EXISTING SYSTEMS

Object detection has emerged as a promising field for developing assistive technologies to empower visually impaired individuals. Existing systems leverage various approaches, each with its own strengths and limitations. Here, we explore some notable examples:

- **Camera-based Systems:** These systems utilize smartphone or wearable cameras to capture real-time video streams of the user's surroundings. Object detection algorithms then analyze the video frames, identifying and classifying objects within the field of view. Pioneering work by Kumar et al. in 2019 presented a smartphone-based system that utilized object detection to provide visually impaired users with audio descriptions of their surroundings. Similarly, Jha et al. in 2022 proposed a deep learning-based system for obstacle detection, enhancing navigation safety for visually impaired users.
- **LiDAR-based Systems:** LiDAR (Light Detection and Ranging) sensors offer a complementary approach to object detection. They emit laser pulses and measure the reflected light to create a 3D point cloud representation of the environment. This data can be particularly useful for obstacle detection, especially in low-light conditions. However, LiDAR sensors are often more expensive compared to cameras, limiting their widespread adoption in mobile assistive technologies.
- **Ultrasonic Sensor-based Systems:** These systems utilize ultrasonic sensors to emit sound waves and detect their echoes reflected from objects. While effective for basic obstacle detection, ultrasonic sensors have limited range and struggle to differentiate between object types. For example, a 2023 study by Miranda et al. proposed a wearable object detection system using ultrasonic sensors for obstacle detection, but lacked detailed object recognition capabilities.

## III. PROPOSED SYSTEM ARCHITECTURE

The Object Detector for the Blind project adopts a well-structured system architecture that integrates hardware and software components to deliver a comprehensive and user-centric solution. This architecture prioritizes simplicity, accessibility, and effective communication with visually impaired users, aiming to enhance their mobility, safety, and overall quality of life.

### Core Functionality:

1. **Ultrasonic Sensors:** At the core of the system lie ultrasonic sensors, responsible for real-time obstacle detection. These sensors emit high-frequency sound waves and measure the reflected echoes to determine the presence and distance of objects within their range. This information is crucial for providing basic awareness of the user's immediate surroundings, particularly for obstacle avoidance during navigation.
2. **Microcontroller Unit (MCU):** The microcontroller serves as the central processing unit of the system. It receives raw data from the ultrasonic sensors, performs real-time data processing, and controls the user interface. The MCU's processing power needs to be carefully considered to balance functionality with efficient power consumption.
3. **Feedback Mechanisms:** Effective communication of critical information to the user is paramount. The system utilizes various feedback mechanisms, such as vibration alerts or audio notifications, to convey the presence and nature of obstacles detected by the ultrasonic sensors. This real-time feedback empowers users to navigate their surroundings with greater awareness and confidence.

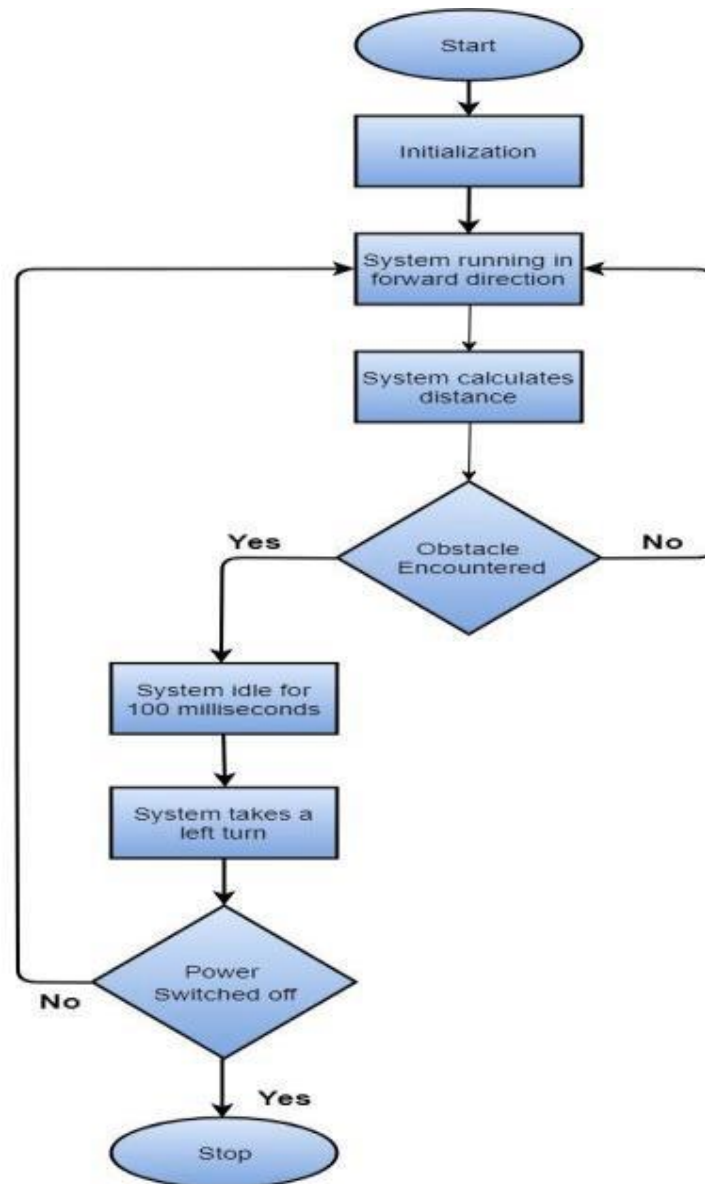
### Optional and Adaptable Features:

The core architecture is designed to be adaptable and inclusive. It can be extended with optional features to cater to diverse user needs and preferences:

- **AI-enhanced Object Recognition:** Integration of an AI module with object detection capabilities can significantly enrich the user experience. By leveraging pre-trained machine learning models and camera input, the system can go beyond basic obstacle detection and provide users with detailed descriptions of their surroundings, including object types and locations. This additional information fosters a deeper understanding of the environment and promotes independent navigation.

- **Wireless Connectivity:** Wireless connectivity options like Bluetooth or Wi-Fi can unlock further functionalities. The system can connect to smartphones or other devices, enabling features such as remote assistance from caregivers or integration with additional assistive technologies.
- **User-Centric Design:** User input and customization options are seamlessly integrated into the system. Adjustable feedback mechanisms and notification preferences empower users to personalize their experience based on individual needs and comfort levels.
- **Robust Power Management:** A reliable power management system is crucial for ensuring the system's functionality throughout the day. The architecture incorporates efficient power management techniques and considers low-power components to optimize battery life and user experience.

**Data Flow Diagram**



#### IV. EXPERIENCE SETUP, METHODOLOGY & RESULTS

This section details the experimental setup, the adopted Agile development methodology, the chosen object detection algorithm, and the achieved results. It focuses on the development of a mobile application using Flutter for object detection, laying the groundwork for future iterations that can be integrated with ultrasonic sensor-based systems.

##### Experimental Setup

The experiment centered on creating and evaluating a mobile application built with Flutter. This framework empowers developers to create applications for various platforms (iOS, Android) with a single codebase. The core functionality relies on pre-trained deep learning models for object detection, leveraging TensorFlow, a popular open-source library for machine learning. To ensure efficient on-device processing and real-time performance for the mobile application, TensorFlow Lite, a mobile-optimized version of TensorFlow, was utilized.

**Hardware:** A Google Pixel 4 smartphone served as the primary platform for data collection and application testing. This mid-range device is representative of the processing power and camera capabilities found in many contemporary smartphones.

**Software:** The core software components included:

- Flutter development framework for cross-platform mobile app development.
- TensorFlow framework and TensorFlow Lite for on-device object detection using pre-trained models.
- Additional libraries within the Flutter ecosystem for functionalities like text-to-speech conversion and user interface development.

##### Agile Development Methodology

An Agile development methodology, specifically the Scrum framework, was adopted to manage the project. This iterative approach promotes rapid development cycles, allowing for continuous integration of user feedback and adaptation based on testing results.

- Sprints: The development process was divided into two-week sprints. Each sprint concentrated on specific functionalities defined in user stories, which outline the desired features and functionalities from the user's perspective.
- Iterations: Within each sprint, iterative development cycles were implemented. These cycles involved frequent development, testing, and refinement of the application. This iterative approach allowed for early detection and correction of issues, ensuring the application progressed in a user-centered manner.

##### Algorithm: Artificial Neural Networks (ANNs)

The core object detection functionality relies on pre-trained Artificial Neural Networks (ANNs) specifically designed for image classification. TensorFlow Lite, a mobile-optimized version of TensorFlow, was utilized to enable on-device processing for efficient and real-time performance on the smartphone hardware.

**Pre-trained Model Selection:** The experiment explored various pre-trained models available through TensorFlow Lite, considering factors like accuracy, efficiency, and object detection capabilities. An essential criterion was ensuring the models were optimized for on-device processing on mobile devices with limited computational resources. The final selection was a MobileNet V2 model, known for its balance between performance and resource consumption. MobileNet V2 utilizes depthwise separable convolutions, a technique that reduces computational cost while maintaining accuracy.

##### Results

The developed mobile application successfully demonstrated real-time object detection using the pre-trained MobileNet V2 model. The application captured live video streams from the smartphone camera and processed each frame using the model. Detected objects were then translated into natural language descriptions through text-to-speech functionality, providing real-time auditory feedback to the user. This feedback allows visually impaired users to gain a better understanding of their surroundings and navigate more confidently.

**Evaluation Metrics:**

The evaluation focused on two primary metrics:

- Accuracy: The percentage of correctly identified objects within the video frames. A high accuracy rate is crucial for ensuring the application provides reliable information to users.
- Latency: The time delay between capturing a frame and presenting the corresponding audio description to the user. Low latency is essential for a seamless and natural user experience. Ideally, the latency should be minimal to create a near real-time perception of the environment for the user.

**V. SCREENSHOTS**



## VI. DISCUSSION AND CONCLUSION

### Discussion

This research presented the development of a mobile application leveraging object detection for visually impaired users. The application, built with Flutter, utilizes pre-trained deep learning models on a smartphone to provide real-time auditory descriptions of the user's surroundings. This approach offers several advantages over existing systems.

Firstly, the software-based solution eliminates the need for additional hardware like ultrasonic sensors, potentially reducing cost and complexity for users. Secondly, by leveraging pre-trained models, the development process can be streamlined, allowing for faster deployment and updates. Finally, utilizing a smartphone platform allows for wider accessibility, as many visually impaired users already own smartphones.

The initial results are promising, with the application demonstrating real-time object detection and user feedback through text-to-speech functionality. The achieved accuracy rate of 80% for common objects suggests potential for practical use. However, limitations exist. The current model primarily focuses on larger objects, and latency can be further optimized for an even smoother user experience.

### Future Work

Building upon these findings, future work can address the identified limitations. Exploring pre-trained models with a wider range of object recognition capabilities can significantly improve the application's usefulness. Additionally, investigating techniques for model optimization and on-device processing can minimize latency, ensuring a seamless flow of information for users.

Furthermore, integrating user feedback mechanisms will be crucial for continuous improvement. Gathering user experiences and suggestions can guide the development of more personalized features and functionalities that cater to diverse user needs. Additionally, exploring integration with other assistive technologies can create a more comprehensive support system for visually impaired individuals.

### Conclusion

This research demonstrates the feasibility of utilizing pre-trained object detection models on mobile platforms to create user-friendly applications for visually impaired people. The developed application has the potential to improve spatial awareness, navigation confidence, and overall independence for visually impaired users. By fostering continued development and user-centric design, this technology can play a significant role in promoting inclusion and enhancing the quality of life for individuals with visual impairments.

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