

# Eco-Eye: Object Detection System for Blind People

Prof. Dipali Mane<sup>1</sup>, Rahul Kolhe<sup>2</sup>, Mohit Patil<sup>2</sup>, Vaishnavi Bharambe<sup>2</sup>, Aarya Raghuvanshi<sup>2</sup>

Assistant Professor, Department of Computer Engineering<sup>1</sup>

Student, Department of Computer Engineering<sup>2</sup>

Alard College of Engineering and Management, Pune, India

**Abstract:** *Visually impaired individuals face significant challenges when it comes to navigating their environment. This project presents a cost-effective and efficient solution for improving mobility through a smart navigation system designed for the visually impaired. Utilizing Raspberry Pi, a camera, and advanced object detection techniques, the system eliminates the need for traditional ultrasonic sensors. By incorporating computer vision, object detection algorithms, and text-to-speech (TTS), the system identifies objects and provides real-time feedback via audio. Furthermore, the system includes an emergency alert feature that allows the user to press a dedicated button to send an email to a guardian, containing their location and a picture of the surroundings. This solution aims to enhance mobility, safety, and independence for visually impaired individuals, while being simple, scalable, and user-friendly. The system is also designed to be affordable and accessible, with future potential for more complex features such as obstacle classification, path guidance, and AI-based personalized navigation.*

**Keywords:** Object Detection, YOLOv3, Raspberry Pi, Assistive Technology, Blind Assistance, COCO Dataset, Edge Detection, Real-Time Processing

## I. INTRODUCTION

According to the World Health Organization (WHO), over 285 million people worldwide experience some form of visual impairment. Among the most significant challenges they face are navigation and obstacle detection, which hinder their independence and safety. Traditional solutions, such as ultrasonic sensor-based sticks or wearable devices, often suffer from issues such as limited accuracy, bulkiness, and a narrow range of detection. Recent advances in computer vision and artificial intelligence, however, offer promising opportunities for developing efficient and lightweight assistive devices. This project introduces a low-cost, camera-based object detection and navigation system powered by a Raspberry Pi, which focuses on real-time detection of objects and obstacles, providing immediate audio feedback using TTS. Additionally, the system enhances user safety with an emergency alert feature, enabling the user to contact a guardian during distress by sending location and visual data. Unlike traditional methods, this approach relies on computer vision for precise and adaptable object detection.

### Aim

The goal of this project is to create an affordable, efficient, and user-friendly object detection and navigation system for visually impaired individuals, utilizing Raspberry Pi and computer vision techniques.

### Scope

- Real-time object detection and audio feedback.
- Emergency email alert system with location and image.
- No reliance on ultrasonic sensors.
- Easily extendable for obstacle avoidance, path finding, and AI-driven navigation.
- Usable in both indoor and outdoor environments



### **Motivation**

The motivation behind this project stems from the need for an affordable, lightweight, and effective navigation solution for the visually impaired. Current technologies often rely on basic obstacle detection via ultrasonic sensors, which have limited range and accuracy. With the rapid advancements in computer vision and the availability of affordable embedded systems like Raspberry Pi, it is now feasible to build a more advanced system that offers real-time object recognition, emergency response capabilities, and improved usability.

## **II. LITERATURE SURVEY**

**Visual-LiDAR Fusion for 3D Object Detection and Tracking on Embedded Devices :** G.-W. Kim proposed a system that combines visual data with LiDAR to enhance 3D object detection and tracking on embedded platforms. By using LiDAR's depth-sensing along with camera-based recognition, their system improves detection accuracy, particularly in low-light environments. However, this fusion approach comes with higher hardware costs and increased computational demands, making it less practical for assistive devices. Despite these limitations, the study highlights the potential of sensor integration to improve detection reliability.

**Real-Time Object Size Measurement Using 3D Cameras :** This research presents a method for determining the size of objects through 3D cameras using depth information. The system provides real-time, accurate measurements without the need for physical contact, making it useful in industries like automation and inspection. While the system's real-time capability and accuracy are its strengths, it is sensitive to camera resolution and environmental factors such as lighting. The study concludes that while 3D cameras offer good measurement precision, improvements in calibration and speed are needed.

**Deep Hash-Based Framework for Object Detection in Remote Sensing :** M. Wang et al. introduced a deep hash-driven network for efficient object detection in remote sensing imagery. This model reduces computational load by compressing data into hash codes, retaining accuracy while speeding up processing. It is especially beneficial for analyzing large volumes of satellite and aerial images. However, the hashing process may result in some information loss, potentially affecting the detection of finer details. The research concludes that this method offers a balance between speed and accuracy, making it effective for large-scale environmental and urban analyses.

**Embedded System for Real-Time Object Detection and Dimension Estimation :** This study discusses an embedded system that performs real-time object detection and size measurement using deep learning techniques. Targeted at resource-constrained platforms, the system is designed to provide quick and reliable results. However, measurement precision is affected by camera calibration and viewing angles, leading to variations in accuracy. The study emphasizes the need for further optimization to improve robustness and consistency in measurements.

### **Contextual Region-Based Scale Adjustment for Object Detection Using Super-Resolution**

K. Akita and N. Ukita proposed a scale- optimized object detection system that adapts to varying image scales. By combining super- resolution techniques with context-aware proposals, the system achieves higher accuracy in recognizing objects of different sizes. However, the use of super-resolution increases computational demands, which may limit real- time applicability. The research suggests that while the method improves detection accuracy, additional optimization is necessary to minimize processing delays.

**Video Object Detection Enhanced Through Blur Analysis :** Y. Wu et al. introduced a technique that incorporates motion blur analysis to improve video-based object detection. By compensating for blur, the system enhances detection accuracy in dynamic environments with fast-moving objects. While effective, the added complexity of blur analysis can cause delays in processing, making real-time deployment more challenging. The study concludes that blur-aware detection works well in motion-intensive scenarios, but optimization is needed for real-time performance.

**Lightweight Multi-Feature Wearable Navigation System for the Visually Impaired** This research proposes a wearable device for the visually impaired that integrates various sensors and computer vision techniques to detect environmental features such as obstacles, slopes, and uneven surfaces. The system also includes functions like location tracking and image capturing to aid navigation. It prioritizes simplicity, ease of use, and comfort while maintaining affordability. User studies indicate that the device improves mobility and safety, offering valuable assistance without overwhelming the user.

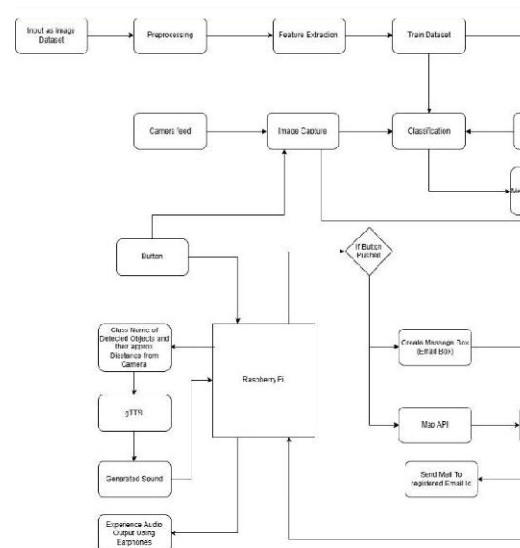


**Obstacle Detection and Distance Estimation for Visually Impaired Navigation** A system focused on detecting obstacles and estimating distances to aid visually impaired navigation, utilizing depth sensors and image processing. This system generates real-time auditory alerts to inform users about nearby obstacles. Its strength lies in the precision of its depth information, though detection accuracy can be affected by environmental noise and lighting conditions. The study suggests that sensor fusion could enhance accuracy and reliability.

**Review of Assistive Technologies for the Visually Impaired :** This paper surveys a range of assistive technologies designed to support visually impaired individuals in daily activities, including object detection systems, navigation aids, and text-to-speech solutions. While these technologies improve independence, challenges like high costs and hardware limitations persist. The study stresses the importance of continuous innovation to overcome these barriers and expand accessibility.

**AI-Driven Assistive Systems for Visually Impaired Individuals :** This research explores modern assistive technologies that incorporate AI and computer vision to assist visually impaired users in real-time. Applications like object recognition and navigation support significantly improve user autonomy. However, the high computational demands and affordability of AI-based systems pose challenges. Despite these hurdles, the study emphasizes the promise of AI in enhancing accessibility, provided research continues to improve efficiency.

### III. SYSTEM ARCHITECTURE



The architecture of the proposed system, Eco-Eye, is designed to assist visually impaired individuals by providing real-time object detection and navigation assistance, while also ensuring user safety through an emergency alert system. The system is powered by a Raspberry Pi, which serves as the central processing unit for all tasks. The primary components include a camera module for capturing images of the surroundings, a speaker or earphones for delivering audio feedback, and a button that triggers an emergency response.

The architecture consists of several interconnected modules. The camera continuously captures live video frames, which are then processed using deep learning algorithms to identify objects within the environment. Object recognition results are sent to a text-to-speech (TTS) engine that converts the information into auditory feedback. For user safety, an emergency button is integrated into the system, which allows the user to send an alert with their location and an image of their surroundings to a pre-configured guardian's email. The entire system operates on the Raspberry Pi, making it both portable and cost-effective, while being sufficiently powerful to process the necessary computations in real time.



The system's modular approach allows for future scalability. Additional sensors, such as infrared or LiDAR, could be added to enhance object detection and depth perception. Similarly, AI-based features such as path finding and more detailed environmental awareness could be integrated into the system, making it adaptable to evolving technological advancements.

#### **IV. SYSTEM COMPONENTS AND WORKING**

The core of the Eco-Eye system consists of several key components, each playing a specific role in ensuring the system functions smoothly. These components include:

- **Raspberry Pi:** The Raspberry Pi is the central processing unit (CPU) of the system. It performs all the data processing, object detection, and system control tasks. Its compact form factor, low power consumption, and computational power make it ideal for use in portable assistive technology.
- **Camera Module:** The camera module captures real-time images of the environment, which are then passed to the Raspberry Pi for processing. The camera's role is crucial for detecting objects and understanding the layout of the user's surroundings.
- **Object Detection:** For object detection, the system uses a pre-trained SSD MobileNet v3 model, which is optimized for running on resource-constrained devices like the Raspberry Pi. This model is capable of detecting various objects such as walls, chairs, tables, and other common obstacles that a visually impaired person may encounter.
- **Text-to-Speech (TTS):** Once objects are detected, the system converts the object labels into audio using the Google Text-to-Speech (gTTS) engine. The user is provided with auditory feedback, which helps them navigate their environment safely and effectively. For instance, if a chair is detected in the path, the system would announce, "Chair ahead, please proceed with caution."
- **Emergency Button:** The system includes a physical emergency button, which, when pressed, sends an immediate alert to a designated guardian's email. The alert includes the current location of the user, as well as an image captured by the camera, which can help the guardian assess the situation.
- **Email Alert System:** In case of an emergency, the system captures a photo of the environment and sends an email to a designated guardian. The email contains the location of the user, derived from GPS coordinates, along with a timestamp and a snapshot of the environment, allowing the guardian to take appropriate action.

The system works as follows:

The camera continuously streams video, which is processed by the Raspberry Pi to detect objects.

Detected objects are identified and categorized using object detection algorithms, with the results being converted into auditory feedback via the TTS engine.

The user can press the emergency button at any time to trigger an immediate alert, which sends the user's location and a photo of their surroundings to the guardian's email.

#### **V. IMPLEMENTATION DETAILS**

The implementation of the Eco-Eye system is divided into several stages, each of which ensures that the system performs optimally within the resource limitations of the Raspberry Pi. The first stage involves setting up the Raspberry Pi and integrating it with the camera module. The camera module is connected to the Pi's camera interface, allowing it to capture real-time video frames.

For object detection, the system uses the SSD MobileNet v3 model, which is trained on the COCO dataset. This model is capable of detecting a wide variety of objects and has been fine-tuned to run efficiently on the Raspberry Pi. To ensure the system operates in real time, the frames captured by the camera are processed as quickly as possible to minimize lag. The object detection process involves passing each frame through the SSD MobileNet v3 model, which outputs the bounding boxes and labels of the detected objects.

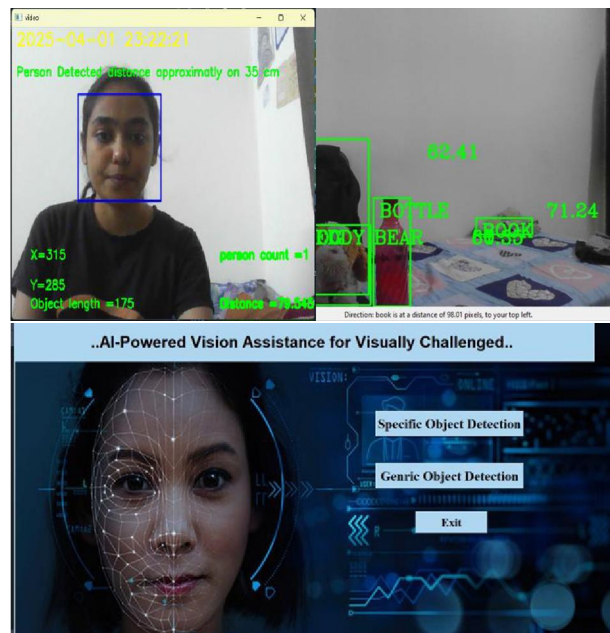


Once the objects are detected, the system uses Google’s TTS (gTTS) API to convert the object labels into speech. The TTS module plays an essential role in delivering feedback to the user. The user hears the system announce the object in the environment, such as “Chair detected on your left” or “Obstacle ahead,” enabling them to make informed decisions while navigating.

The emergency alert system is integrated through a simple button press. The user presses the button, which triggers a script to capture an image from the camera and send an email containing the image and location data. The email is sent using Python’s smtplib library, and the system uses GPS coordinates obtained from a connected GPS module or a Wi-Fi-based location service. The email contains not only the user’s location but also a snapshot of the surrounding area, which is invaluable for guardians in case of an emergency.

The system’s overall design emphasizes efficiency, real-time performance, and ease of use. The combination of Raspberry Pi and camera-based object detection ensures that the system is lightweight and portable, while the TTS engine allows for seamless interaction. Moreover, the emergency feature adds an extra layer of safety, ensuring that the user can easily communicate their location to a guardian when needed.

The final system is housed in a lightweight, durable enclosure, making it easy to wear or carry. It is optimized for minimal power consumption, ensuring that it can run for extended periods without the need for frequent recharging. As the system relies on open-source technologies such as the Raspberry Pi and gTTS, it remains affordable and accessible for a wide range of users.



**VI. CONCLUSION**

In conclusion, the integration of Object Detection for visually impaired individuals, utilizing machine learning frameworks like TensorFlow, offers a promising solution to enhance independence and mobility for the blind. The combination of innovative technologies like deep learning and object detection, integrated with hardware such as Raspberry Pi, cameras, and sensors, has the potential to create smart navigation systems that can significantly assist individuals in real-time. These systems are equipped with features such as alert notifications to guardians, the ability to detect objects and obstacles, and text-to-speech functionalities, which collectively make it possible for blind people to move through their environments more safely and autonomously.

However, the implementation of these systems does present challenges, especially when it comes to optimizing accuracy in object detection under different lighting conditions, handling varying objects in dynamic environments, and ensuring the reliability of the sensors. Additionally, as these technologies continue to evolve, there are opportunities for further





enhancing the user experience, such as incorporating more sophisticated machine learning models for better recognition or integrating advanced sensors for greater environmental awareness.

With the current advancements in artificial intelligence, this project opens the door to more inclusive, accessible technologies that could transform the daily lives of visually impaired individuals. Further research and development could lead to even more intuitive and effective solutions, making it possible for individuals with visual impairments to navigate the world with greater ease, freedom, and confidence.

#### REFERENCES

- [1]. M. Sualeh, G.-W. Kim: Visual-LiDAR Based 3D Object Detection and Tracking for Embedded Systems VOLUME 8, 2020
- [2]. Automatic Method for Measuring Object Size Using 3D Camera Cuong Vo-Le, Pham Van Muoi, Nguyen Hong Son, Nguyen Van San, Vu Khac Duong and Nguyen Thi Huyen
- [3]. M. Wang et al.: Deep Hash Assisted Network for Object Detection in Remote Sensing Images VOLUME 8, 2020
- [4]. An Embedded Real-Time Object Detection and Measurement of its Size Nashwan Adnan OTHMAN, Mehmet Umut SALUR, Mehmet KARAKOSE, Ilhan AYDIN
- [5]. K. Akita, N. Ukita: Context-Aware RDSPs for Scale-Optimized Object Detection Using SR VOLUME 11, 2023
- [6]. Y. WU et al.: Video Object Detection Guided by Object Blur Evaluation VOLUME 8, 2020
- [7]. Design, development and performance analysis of cognitive assisting aid with multi sensor fused navigation for visually impaired people, Myneni Madhu Bala<sup>1</sup>, D. N. Vasundhara<sup>2\*</sup>, Akkineni Haritha<sup>3</sup> and CH. V. K. N. S. N. Moorthy<sup>4,5\*</sup>, Bala et al. Journal of Big Data (2023)10:21
- [8]. X. Leong, R. Kanesaraj Ramasamy: Obstacle Detection and Distance Estimation for Visually Impaired People VOLUME 11, 2023
- [9]. Assistive Technology for the Visually impaired using Computer Vision, Megha P Arakeri, Tazeen Munnavar, Anusha Sankar, Madhura M, Keerthana N S, Information Science and Engineering Ramaiah Institute of Technology
- [10]. Ghosh, Amit; Al Mahmud, Shamsul Arefeen; Uday, Thajid Ibna Rouf; Farid, Dewan Md.: Assistive Technology for Visually Impaired using Tensor Flow Object Detection in Raspberry PI and Coral USB Accelerator, Proceedings of the 2020 IEEE Region 10 Symposium, TENSYP 2020

