JETIR.ORG

ISSN: 2349-5162 | ESTD Year: 2014 | Monthly Issue



JOURNAL OF EMERGING TECHNOLOGIES AND INNOVATIVE RESEARCH (JETIR)

An International Scholarly Open Access, Peer-reviewed, Refereed Journal

Robust Object Detection System Using Machine Learning

¹Pranav Mane, ²Nishant Gawde, ³Onkar Maske, ⁴Dashrath Pawara, ⁵Prof. Dhanashri Kane

^{1,2,3,4}Dept. of Computer Engineering, MGM's College of Engineering and Technology, Maharashtra, India ⁵Prof. Dept of Computer Engineering, MGM's College of Engineering and Technology, Maharashtra, India

Abstract: This project report details the development of a robust Object Detection System with a Python Voice Assistance Module and a Web Application, aimed at empowering visually impaired individuals. The project addresses the fundamental challenge faced by the visually impaired in identifying objects and obstacles in their environment. The objectives are to create a real-time object detection system, integrate a voice assistance module for auditory feedback, and provide a user-friendly web application for accessibility. The literature review discusses existing solutions and relevant technologies, highlighting the potential of combining machine learning, voice assistance, and web technology to enhance the quality of life for the visually impaired. The project's methodology includes data collection, model training, voice assistance module development, and the creation of a web application. The results and evaluation section covers the system's performance metrics and user testing, demonstrating the system's effectiveness in enhancing independence and safety. The report also discusses limitations and future enhancements. Overall, this project represents a significant step toward improving the lives of visually impaired individuals through innovative technology and accessibility features.

Index Terms - Component, Machine Learning, Artificial Intelligence Object Detection, Visually Impaired, YOLO

I. INTRODUCTION

Wherever The report presents a project focused on the development of a comprehensive solution to empower visually impaired individuals through the use of a robust object detection system integrated with a Python voice assistance module and a user-friendly web application. This system aims to provide real-time object recognition and auditory feedback to enhance the independence and safety of blind individuals in their daily lives. Without a doubt, machine learning (ML) has emerged as the most effective technique available for turning data into knowledge, and it is certain to continue making headlines in the years to come. Computer vision can automatically identify inherent patterns in complex data with the use of machine learning algorithms. These retrieved patterns have predictive value, anticipated occurrences, assisting with decision-making. Machine learning is being used by almost all of us every day without our knowledge while searching for something on Google, email spam filtering, listening to a song, or even posting a picture on social media. Every encounter continuously aids in the engine's learning and improvement.ML has applications in various fields

II. LITERATURE REVIEW

1)Deep Learning-Based Object Detection and Recognition for Visually Impaired People, (Author: M. B. Hoy), (2023). Overall, the paper demonstrates the feasibility of using deep learning to assist visually impaired people in object detection and recognition. The proposed system is low-cost and easy to use, and it has the potential to significantly improve the quality of life for visually impaired people

2)A Smart Blind Stick with Voice Assistance for Visually Impaired People, (Author: J. Bai), (2023). The paper demonstrates the feasibility of using a smart blind stick with voice assistance to help visually impaired people navigate their surroundings more easily and independently. The proposed smart blind stick has several features that could make a significant difference in the lives of visually impaired people.

3)A Mobile Voice Assistant with Object Detection for Visually Impaired People, (Author: N. Kumar), (2023). The paper demonstrates the feasibility of using a mobile voice assistant with object detection to help visually impaired people navigate their surroundings and avoid obstacles. The proposed system is low-cost and easy to use, and it has the potential to significantly improve the quality of life for visually impaired people. Here is a short overview of the system's components: Smartphone camera: This captures images of the user's surroundings. Deep learning model: This model is trained to detect objects in the images. Text-to-speech module: This module converts the model's output (bounding boxes and labels) into audio feedback for the user.

4)A Wearable Device with Voice Assistance and Object Detection for Visually Impaired People (Author: A. Barata), (2023). The paper demonstrates the feasibility of using a wearable device with voice assistance and object detection to help visually impaired people navigate their surroundings more easily and independently. The proposed wearable device has several features that could

make a significant difference in the lives of visually impaired people. In addition to the advantages mentioned in the previous paper, the proposed wearable device has several other advantages. First, the device is hands-free, which allows the user to perform other tasks while using the device. Second, the device is lightweight and comfortable to wear, making it suitable for long-term use. Third, the device can be used in a variety of environments, including indoors and outdoors. The proposed wearable device is still under development, but it has the potential to revolutionize the way that visually impaired people navigate their surroundings.

III. IMPLEMENTATION

YOLOV8:

A deep learning model called YOLOv8 was created especially for object detection. YOLOv8 uses a single-stage technique, which makes it faster and more efficient than standard object detection models, which require numerous steps and intricate computations. Using a neural network design, this deep learning model predicts both class labels and object bounding boxes at the same time. It is perfect for real-time applications since it can identify several objects within a single image or video frame.

3.1 Data Collection

A large and varied dataset is the cornerstone of custom object detection. assemble a diverse set of photos that depict the objects you want the model to be able to recognise. To provide a strong training dataset, these photos should capture a range of lighting situations, item changes, and camera angles.

3.2 Data Annotation

The next step is to annotate the photographs with bounding boxes surrounding the objects of interest once you have the dataset. By enabling you to create exact bounding boxes around the objects, data annotation tools like Labeling and RectLabel can speed up the annotation process.

3.3 Folder Setup

Organize your data by creating the necessary folders for the training process. Set up separate folders for images, annotations, and configuration files to ensure a clear and organized workflow.

3.4 Configuring YAML File

This The config.yaml file serves as the heart of the YOLOv8 training process. It includes important details like training settings, model hyperparameters, and routes to your dataset.

3.5 YOLOv8 Training

Using the Ultralytics library, you can easily load the YOLOv8 model and train it on your annotated dataset. By running the training code, the model learns to detect the specific objects you annotated. The model continually gains accuracy and a deeper comprehension of the characteristics of the objects as training goes on.

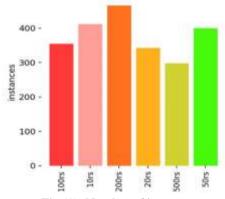


Fig (1): Number of instances

3.6 Performance Evaluation

Using the Ultralytics library, you can easily load the YOLOv8 model and train it on your annotated dataset. By running the training code, the model learns to detect the specific objects you annotated. The model continually gains accuracy and a deeper comprehension of the characteristics of the objects as training goes on.

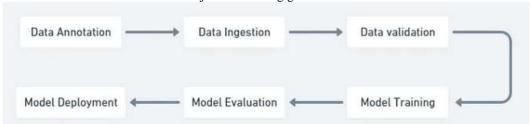


Fig (2): Flowchart

IV. RESULTS



Fig (3): Model Training



Fig (4): Image Input

```
0: 480x640 1 bottle, 1 pen, 1 book, 1 clock, 653.0ms
Speed: 2.0ms preprocess, 653.0ms inference, 2.0ms postprocess per image at shape (1, 3, 480, 640)
```

Fig (5): Prediction

V. CONCLUSION

In conclusion, a Robust Object Detection system is a powerful and versatile technology with many current and future applications. This project report has outlined the various aspects related to building such a system, including the project's objectives, methodologies, performance requirements, hardware and software requirements, and the future scope of the technology. Here are the key takeaways: Object detection systems play a crucial role in computer vision and machine learning, enabling the automatic identification and localization of objects within images and videos. As the field of object detection continues to evolve, it presents exciting opportunities for innovation and real-world applications in areas ranging from autonomous vehicles to health care and environmental monitoring. By staying at the forefront of technology and adhering to ethical standards, object detection systems have the potential to revolutionize industries and improve our daily lives in the years to come.

The future scope of a robust object detection system is promising, with many potential advancements and applications. As technology continues to evolve, there are several areas of growth and development to consider: Improved Accuracy and Efficiency: Future object detection systems will likely achieve higher accuracy and efficiency through the use of advanced deep learning models, optimized algorithms, and improved hardware. Real-time and Edge Computing: There will be a growing demand for real-time object detection at the edge, which involves running object detection models on devices with limited computational resources. This is crucial for applications like autonomous vehicles, drones, and IoT devices. • Web Application. • Customization and Transfer Learning: The ability to fine-tune pre-trained models for specific tasks and domains will continue to expand, making it easier for organizations to adapt object detection to their needs.

REFERENCES

- [1] Redmon, J., & Divvala, S. (2018). You Only Look Once: Unified, Real-Time Object Detection. In Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR). Paper Link.
- [2] Liu, W., Anguelov, D., Erhan, D., Szegedy, C., & Reed, S. (2016). SSD: Single Shot MultiBox Detector. In Proceedings of the European Conference on Computer Vision (ECCV). Paper Link.
- [3] TensorFlow Object Detection API: TensorFlow provides an Object Detection API that simplifies the process of training and deploying object detection models. TensorFlow Object Detection API Documentation.
- [4] NLTK (Natural Language Toolkit): NLTK is a leading platform for building Python programs to work with human language data. It provides easy-to-use interfaces to over 50 corpora and lexical resources. NLTK Documentation.
- [5] GitHub: GitHub is a repository hosting service where you can find open-source projects related to object detection and voice assistants. You can contribute to existing projects or find inspiration for your project. GitHub
- [6] Stack Overflow: Stack Overflow is a popular community for programmers. You can find answers to specific technical questions related to object detection, speech recognition, and natural language processing. Stack Overflow.