

DRONE OBJECT DETECTION MODELS FOR HIGHLY RESTRICTED AREAS

Tarun Gowda S D¹, Dr. T Vijaya Kumar²

Student, Department of MCA, Bangalore Institute of Technology, Karnataka, India¹

Professor & Head, Department of MCA, Bangalore Institute of Technology, Karnataka, India²

Abstract: Unmanned Aerial Vehicles (UAVs), colloquially referred to as drones, have witnessed exponential growth in their usage across diverse domains, ranging from agriculture and infrastructure inspection to surveillance and cinematography. This surge in popularity has been facilitated by advancements in drone technology, making them more accessible, affordable, and versatile. However, along with their myriad benefits, drones also pose challenges, particularly in areas such as privacy, security, and safety. As such, the development of robust drone detection systems becomes imperative to address these concerns and ensure responsible drone usage.

Keywords: DCSASS Dataset, Deep learning algorithms, ResNet50, I3D.

I. INTRODUCTION

Unmanned Aerial Vehicles (UAVs), commonly known as drones, have become increasingly prevalent in various fields, presenting both opportunities and challenges. This project Promotes the creation of a drone object system detection using the YOLOv8 model. The dataset comprising drone images was collected from Roboflow and annotated using PASCAL VOC XML format. The dataset was then divided into training, testing, and validation sets to facilitate robust model training. The YOLOv8 model was trained on the annotated dataset, optimizing parameters to minimize loss functions. Evaluation metrics such as precision, recall, and mAP were utilized to assess the model's performance, demonstrating high accuracy in detecting drones.

Furthermore, a Flask web application was developed to provide a user-friendly interface for utilizing the trained model. This application accepts both images and videos as input, enabling real-time drone detection. This project presents a comprehensive approach to drone object detection, encompassing data collection, model training, evaluation, application development, and hardware integration. The successful implementation of the system demonstrates its potential for various applications, including security, surveillance, and safety in areas where drone presence needs monitoring and control.

Keywords: Unmanned Aerial Vehicles (UAVs), YOLOv8 model, Roboflow, PASCAL VOC XML format, User-friendly interface.

II. LITERATURE SURVEY

Aswini N. [1] presents a study on drone object detection using deep learning algorithms. The research aims to meet the increasing demand for precise and efficient object detection in drone imagery. The authors explore the application of deep learning techniques, which have demonstrated success in various computer vision tasks, including object detection. By utilizing convolutional neural networks (CNNs) specifically designed for object detection, the authors propose an innovative method for detecting objects in drone-captured images.

R. Jadhav, R. Patil. [2] presents a study on drone-based object detection using artificial intelligence (AI) techniques. The authors aim to a develop an intelligent system capable of autonomously detecting objects from drone-captured images. The study likely explores the integration of AI algorithms, including machine learning and the deep learning techniques, to analyze visual data obtained from drones.

W. Budiharto, A. S. [3] examines rapid object detection for quadcopter drones using deep learning techniques. The research addresses the need for efficient, real-time object detection in applications such as surveillance, search and rescue, and aerial monitoring. The authors explore deep learning models specifically designed for resource-constrained environments like drones, aiming to achieve fast and accurate object detection.

T. Abdellatif, M. A. [4] the authors introduce DroMOD, a drone-based multi-scope object detection system. The research focuses on developing a comprehensive solution for object detection tasks using drones across various scenarios and applications.

M. Issame and A. [5] the authors present a study on real-time object detection with drones using deep learning algorithms. The research aims to develop an efficient and accurate solution for detecting objects from drone-captured images in real-time scenarios. The authors likely explore the application of deep learning models, such as a conventional neural networks (CNNs) or recurrent neural networks (RNNs), optimized for real-time performance and accuracy in object detection tasks.

Keywords: Convolutional neural networks (CNN), Human Activity Recognition, Recurrent Convolutional Attention Network, Multi-class activity recognition.

III. EXISTING SYSTEM

Existing drone detection systems typically rely on traditional computer vision algorithms or simplistic machine learning models, which often exhibit limitations in terms of accuracy and speed. These systems may struggle to detect drones in cluttered or dynamic environments, leading to high false positive or false negative rates. Moreover, many existing solutions lack the flexibility to adapt to evolving drone technologies and tactics, rendering them less effective in real-world scenarios.

IV. PROPOSED SYSTEM

The proposed drone object detection system utilizes the YOLOv8 model, a state-of-the-art deep learning algorithm renowned for its real-time performance and high accuracy. By training the model on annotated drone datasets and optimizing its parameters, the system aims to achieve superior detection capabilities across diverse environments and conditions. Additionally, the system incorporates a Flask web application interface for seamless interaction, allowing users to upload images or videos for drone detection.

V. IMPLEMENTATIONS AND METHODOLOGY

Software Architecture:

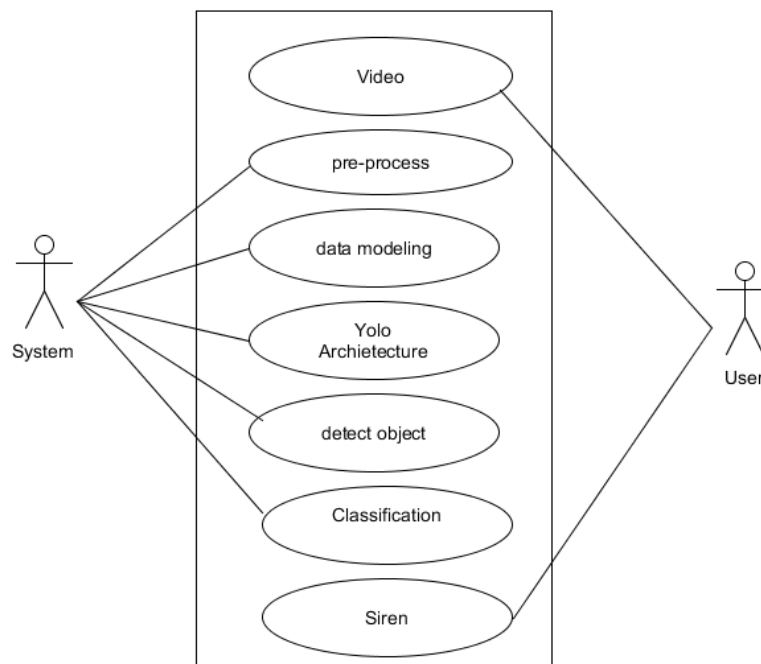


Fig-1: Software Architecture

The project is implemented using python which is an object oriented programming language and procedure oriented programming language. Object oriented programming is an approach that provides a way of modularizing program by creating partitioned memory area of both data and function that can be used as a template for creating copies of such module on demand.

This project is implemented using python programming language. Python is dynamically typed and garbage-collected. It supports multiple programming paradigms, including procedural, object-oriented, and functional programming. Python is often described as a "batteries included" language due to its comprehensive standard library. The machine Learning techniques are used in this project. Implementation of software refers to the final installation of the package in its real environment, to the satisfaction of the intended users and the operation of the system. The people are not sure that the software is meant to make their job easier.

VI. RESULTS

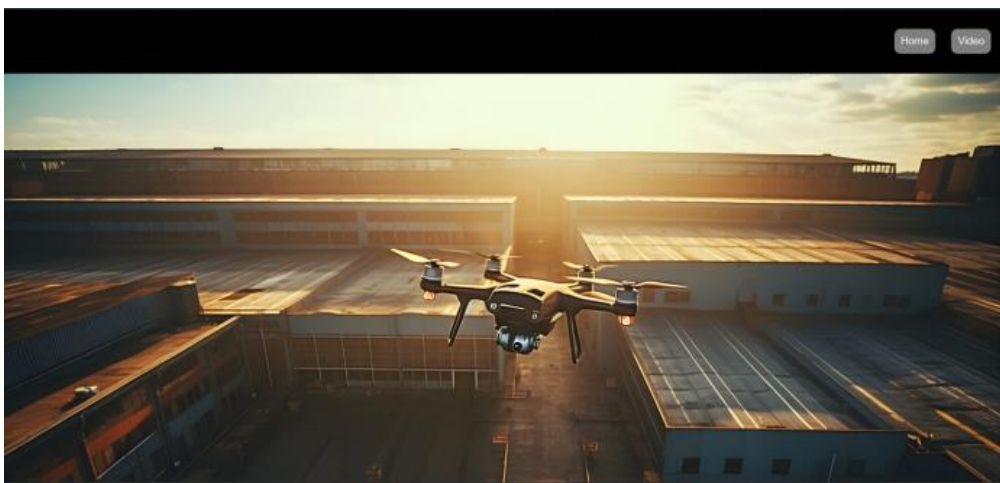


Fig-2: Home Page

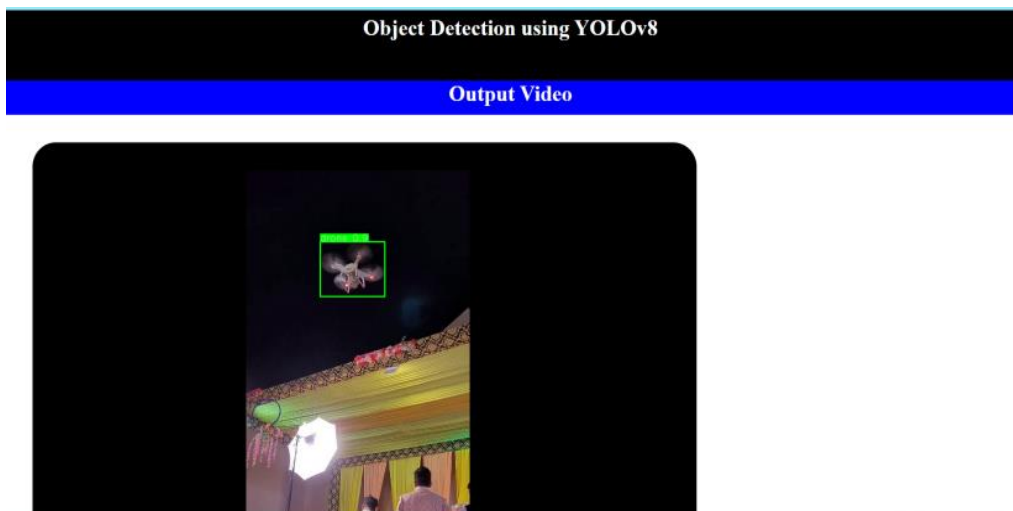


Fig-3: Output Page

VII. CONCLUSION

The development of the web application for drone object detection using YOLOv8 for real-time detection marks a significant advancement in automated surveillance and security systems. The project has not only met its objectives but has also showcased promising accuracy rates and practical utility in various applications, such as perimeter security, disaster management, and environmental monitoring.

The project's success is underscored by its high accuracy rates in object detection, achieved through rigorous model training and optimization processes. By leveraging state-of-the-art machine learning techniques and datasets collected from drone-captured images and videos, the YOLOv8 model demonstrates robust performance in accurately identifying and classifying objects of interest in real-world scenarios. Evaluation metrics such as precision, recall, and mean Average Precision (mAP) have been used to quantify the model's accuracy, ensuring reliable detection results and minimizing false positives and false negatives. This real-time alerting mechanism not only enhances situational awareness but also facilitates proactive security measures and incident mitigation strategies. Looking ahead, there are opportunities for further enhancing the accuracy and functionality of the system. This includes ongoing refinement of the object detection model through continuous training and fine-tuning, as well as exploration of advanced techniques such as transfer learning and ensemble methods to improve detection performance across diverse environmental conditions and object types. Additionally, integrating feedback mechanisms and user interactions into the system can enable adaptive learning and customization, enhancing its adaptability to evolving security requirements and user preferences.

VIII. FUTURE ENHANCEMENTS

In considering future enhancements to the web application for drone object detection, several avenues present themselves for further development and refinement. One key area for improvement lies in enhancing the accuracy and robustness of the object detection model. This could involve exploring advanced machine learning techniques, such as ensemble learning, semi-supervised learning, or attention mechanisms, to improve detection performance across various environmental conditions, object scales, and orientations. Additionally, ongoing data collection and annotation efforts can help expand the diversity and size of the training datasets, enabling the model to better generalize to new scenarios and object classes. Another potential enhancement involves incorporating advanced object tracking and recognition capabilities into the system. By integrating algorithms for object tracking, such as Kalman filters or Hungarian algorithms, the system can track detected objects over time and across multiple frames, providing valuable insights into object trajectories and behavior patterns. Furthermore, integrating recognition algorithms, such as facial recognition or license plate recognition, can extend the system's applicability to specific use cases, such as identifying individuals or vehicles of interest within the monitored area.

REFERENCES

- [1]. Aswini N , "Drone Object Detection Using Deep Learning Algorithms," 2021 Third International Conference on Inventive Research in Computing Applications (ICIRCA), Coimbatore, India, 2021, pp. 1187-1192, doi: 10.1109/ICIRCA51532.2021.9544983.
- [2]. R. Jadhav, R. Patil, A. Diwan, S. M. Rathod and A. Sharma, "Drone Based Object Detection using AI," 2022 International Conference on Signal and Information Processing (IconSIP), Pune, India, 2022, pp. 1-5, doi: 10.1109/ICoNSIP49665.2022.10007476.
- [3]. W. Budiharto, A. A. S. Gunawan, J. S. Suroso, A. Chowanda, A. Patrik and G. Utama, "Fast Object Detection for Quadcopter Drone Using Deep Learning," 2018 3rd International Conference on Computer and Communication Systems (ICCCS), Nagoya, Japan, 2018, pp. 192-195, doi: 10.1109/CCOMS.2018.8463284.
- [4]. T. Abdellatif, M. A. Sedrine and Y. Gacha, "DroMOD: A Drone-Based Multi-Scope Object Detection System," in *IEEE Access*, vol. 11, pp. 26652-26666, 2023, doi: 10.1109/ACCESS.2023.3253767.
- [5]. M. Issame and A. Benyounes, "Real Time Object Detection With Drone Using Deep Learning Algorithm," 2022 International Conference of Advanced Technology in Electronic and Electrical Engineering (ICATEEE), M'sila, Algeria, 2022, pp. 1-6, doi: 10.1109/ICATEEE57445.2022.10093104.
- [6]. Y. Zhang, L. Shen, X. Wang and H. -M. Hu, "Drone Video Object Detection using Convolutional Neural Networks with Time Domain Motion Features," 2020 IEEE Conference on Multimedia Information Processing and Retrieval (MIPR), Shenzhen, China, 2020, pp. 153-156, doi: 10.1109/MIPR49039.2020.00039.
- [7]. D. Pietrow and J. Matuszewski, "Objects detection and recognition system using artificial neural networks and drones," 2017 Signal Processing Symposium (SPSymo), Jachranka, Poland, 2017, pp. 1-5, doi: 10.1109/SPS.2017.8053689.
- [8]. Q. Wu and Y. Zhou, "Real-Time Object Detection Based on Unmanned Aerial Vehicle," 2019 IEEE 8th Data Driven Control and Learning Systems Conference (DDCLS), Dali, China, 2019, pp. 574-579, doi: 10.1109/DDCLS.2019.8908984.
- [9]. O. Sahin and S. Ozer, "YOLODrone: Improved YOLO Architecture for Object Detection in Drone Images," 2021 44th International Conference on Telecommunications and Signal Processing (TSP), Brno, Czech Republic, 2021, pp. 361-365, doi: 10.1109/TSP52935.2021.9522653.