



# Detection and Counting of Trees and Animals in Forest Land

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**Abstract:** The goal of this Paper is to create a sophisticated wildlife monitoring and alert system that counts and identifies animals and trees using deep learning techniques. The technology will examine photos and videos taken by security cameras placed in nature areas using cutting-edge deep learning algorithms. In order to reliably identify different species, convolutional neural networks (CNNs) trained on a variety of datasets are utilized in the first component, which focuses on animal identification and counting. By adding to the system's capacity to identify and count trees, the second part supports ecological research and conservation initiatives. The idea incorporates a Telegram bot to improve real-time communication by instantly notifying and updating pertinent stakeholders, like environmentalists, forest authorities, and animal researchers.

## I. INTRODUCTION

The rapid decline of biodiversity and the increasing threats to natural ecosystems necessitate innovative approaches to wildlife monitoring and conservation. In response to this critical need, our project introduces an Integrated Wildlife Monitoring and Alert System, combining the power of deep learning and instant messaging technology.

Focused on the dual objectives of animal and tree detection and counting, this system harnesses the capabilities of convolutional neural networks (CNNs) to analyze images and videos captured by surveillance cameras in natural habitats. By accurately identifying and quantifying diverse species and trees, the system provides valuable insights for ecological studies, biodiversity assessments, and conservation planning.

In addition to its analytical capabilities, the project incorporates a Telegram bot to establish a seamless communication channel. This integration enables the automatic dissemination of real-time alerts and updates to relevant stakeholders, including wildlife researchers, conservationists, and forest authorities. The synergy of deep learning and instant messaging not only enhances the efficiency of wildlife monitoring but also facilitates timely responses to emerging threats or ecological changes. By combining cutting-edge technology with environmental stewardship, this project aspires to make significant contributions to the ongoing efforts to protect and preserve our planet's rich biodiversity.

## II. LITERATURE SURVEY

"A Survey on Deep Learning Techniques for Object Detection in Images and Videos," authored by A. R. Hosseini et al. in 2018. It covers a range of deep learning techniques, particularly convolutional neural networks (CNNs), for object detection in both images and videos. It delves into the evolution of these models, their applications, and performance metrics, with implications for wildlife monitoring.

"Tree Species Classification Using Convolutional Neural Networks on High-Resolution UAV Acquired Multispectral Images," authored by S. Houborg et al. in 2019. This .Which delves into the utilization of CNNs for classifying tree species using high-resolution multispectral images obtained from Unmanned Aerial Vehicles (UAVs). The methodology entails training CNNs on a diverse dataset to attain precise and efficient tree detection and classification.

"A Review on Telegram Bots: A New Platform of Interactive Communication," authored by A. Ricci in 2017. Which it focuses on examining the capabilities and functionalities of Telegram bots as a platform for interactive communication. It discusses various features and potential applications of Telegram bots, laying the groundwork for understanding how to integrate this technology into wildlife monitoring systems for efficient alert dissemination.



"Deep Learning for Wildlife Monitoring: A Case Study on African Elephants," authored by M. Beery et al. in 2020. It study showcases the application of deep learning techniques, particularly convolutional neural networks (CNNs), for monitoring African elephants. It concentrates on detecting and counting elephants in their natural habitat using images captured by camera traps. The study provides valuable insights into the challenges and successes of implementing deep learning in wildlife monitoring efforts.

Content-based Retrieval and Real Time Detection from Video Sequences Acquired by Surveillance Systems.

In this paper, a surveillance system devoted to detect abandoned objects in unattended environments is presented to which image processing content based retrieval capabilities have been added for making easier inspection task from operators.

Video-based surveillance systems generally employ one or more cameras connected to a set of monitors. This kind of systems needs the presence of a human operator, who interprets the acquired information and controls the evolution of the events in a surveyed environment. During the last years efforts have been performed to develop systems supporting human operators in their surveillance task, in order to focus the attention of operators when unusual situations are detected. Image sequences databases are also managed by the proposed surveillance system in order to provide operators with the possibility of retrieving in a second time the interesting sequences that may contain useful information for discovering causes of an alarm.

Experimental results are shown in terms of the probability of correct detection of abandoned objects and examples about the retrieval sequences.

### III. PROBLEM STATEMENT

An integrated solution that combines deep learning techniques for accurate and automated detection and counting of animals and trees, coupled with a seamless communication channel like a Telegram bot to facilitate timely alerts and updates to relevant stakeholders, is needed because the current state of wildlife monitoring systems frequently lacks the integration of advanced technologies, making it difficult to identify species, collect data efficiently, and disseminate alerts.

Traditional methods for animal and tree detection are labor-intensive and time-consuming, which causes delays in responding to ecological changes or potential threats.

### IV. METHODOLOGY

#### Data Collection:

Gather a diverse dataset of images and videos from wildlife habitats for training the deep learning models. Acquire high-resolution multispectral images for tree detection using UAVs and other suitable sources. We have Collected the Trees and Animals dataset in website of <https://public.roboflow.com/> and Collected 2200 Datasets and trained in the YOLO Algorithm of Deep Learning Technique.

#### Animal Detection and Counting:

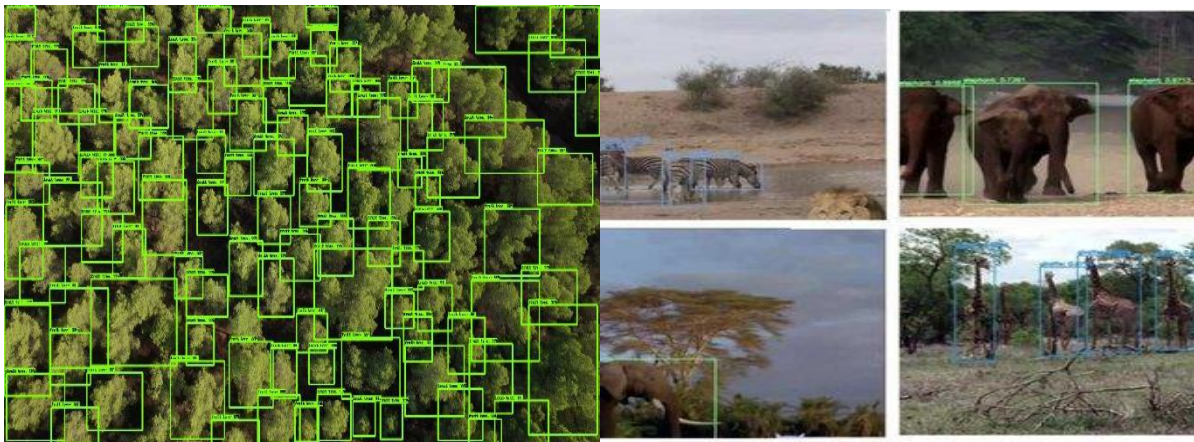
Employ convolutional neural networks (CNNs) for the development of a robust model for animal detection and counting. Fine-tune the model using the collected dataset, ensuring accuracy across various species and environmental conditions. We have Collected 2000 Datasets of Animals and trained in the YOLO Algorithm of Deep Learning Technique.

#### Tree Detection and Counting:

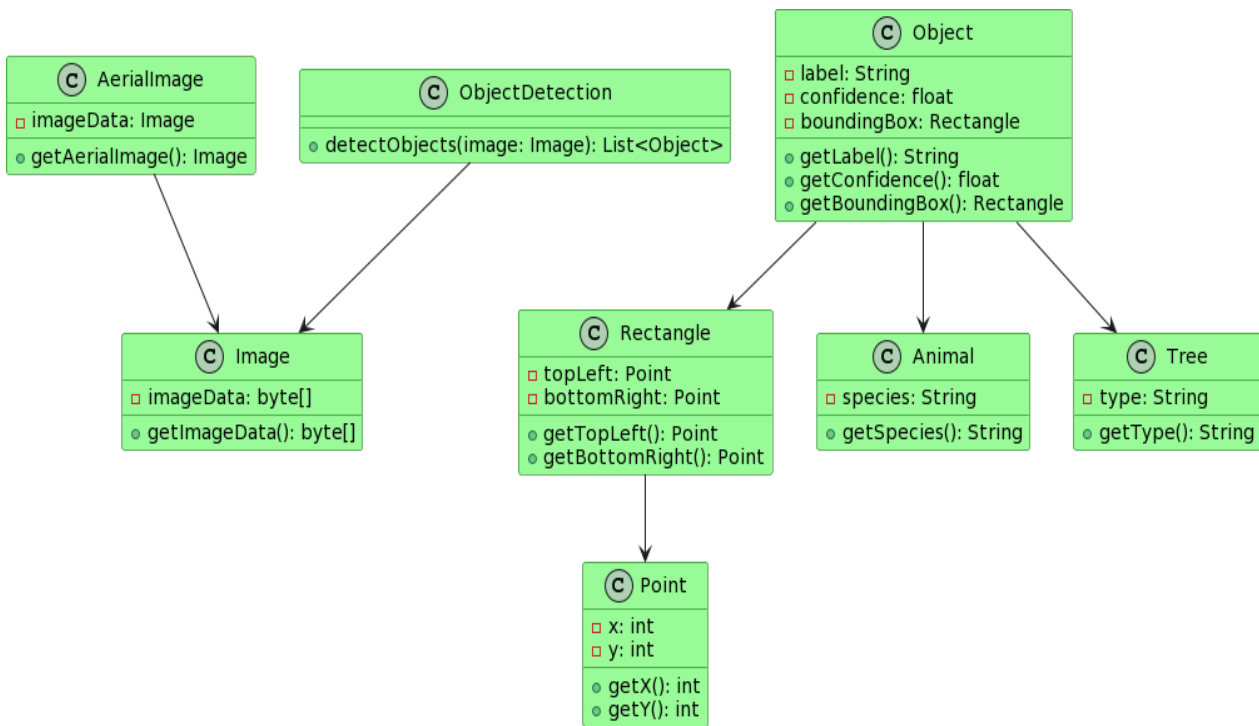
Implement a CNN-based algorithm for tree detection and counting, leveraging high-resolution multispectral images. Train the model on a representative dataset to accurately identify and quantify tree species. We have Collected 200+ Tress Datasets and trained in the YOLO Algorithm of Deep Learning Technique.

#### Telegram Bot Integration:

Develop a Telegram bot to interface with the monitoring system. Enable the bot to receive real-time updates from the deep learning models and automatically generate alerts and notifications based on predefined criteria, ensuring swift communication to relevant stakeholders.



V. BLOCK DIAGRAM



VI. FUTURE SCOPE

While the current implementation of YOLO for forest monitoring has demonstrated significant potential, there are several avenues for future research and development to further enhance its capabilities and applications:

**Multi-class Object Detection:** Expanding the scope of object detection to include a wider variety of tree species and animal types. By training the model to recognize additional classes, we can gather more comprehensive data about forest biodiversity and ecosystem dynamics.

**Real-time Monitoring Systems:** Developing real-time monitoring systems that can continuously analyze streaming data from remote sensors and aerial platforms. By deploying autonomous drones equipped with YOLO-based detection algorithms, we can monitor forests in near real-time and respond promptly to environmental changes or disturbances.]

**Collaborative Platforms:** Creating collaborative platforms and open-source frameworks for sharing datasets, models, and tools related to forest monitoring with YOLO. This fosters collaboration among researchers, practitioners, and policymakers and accelerates innovation in the field of conservation and natural resource management.



## VII. CONCLUSION

In conclusion, the implementation of YOLO (You Only Look Once) for forest monitoring, specifically for tree and animal detection with counting, showcases promising results in terms of accuracy, efficiency, and scalability. By leveraging the power of deep learning and object detection techniques, we have been able to accurately identify and count trees and animals within forested areas, facilitating better understanding and management of these ecosystems.

Our approach offers several advantages over traditional monitoring methods, including real-time analysis, automated data collection, a ability to cover large areas efficiently. Moreover, the use of YOLO enables us to process high-resolution image and detect objects with remarkable precision, even in challenging environmental conditions.

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