



**WEED DETECTION USING DEEP LEARNING AND IMAGE PROCESSING**

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**Abstract:**

Weed identification in vegetable plantations poses a significant challenge due to the variable plant spacing. While traditional crop weed identification methods focus on direct detection of weeds, this approach is not effective due to the wide variety of weed species. In this paper, we propose a new method that blends deep learning and image processing technologies to tackle this challenge. We trained a Center Net model to detect veggies and create bounding boxes around them, allowing us to label the remaining green objects as weeds. By concentrating solely on detecting vegetables, our model avoids the complexity of handling numerous weed species. Our approach provides a promising solution for weed identification in vegetable plantations and paves the way for further research in this field

**I. INTRODUCTION**

Weed identification in vegetable plantations has always been a significant challenge due to the variability of plant spacing and the wide variety of weed species. Traditional crop weed identification methods have focused on direct detection of weeds, but these approaches have proven ineffective in identifying weeds in vegetable plantations. The sheer number of weed species and the close proximity of crops make it challenging to identify and remove weeds without damaging the crops. In this paper, we propose a new method for weed identification in vegetable plantations that combines deep learning and image processing technologies. Our approach utilizes a Center Net model that detects vegetables in the plantation and creates bounding boxes around them, leaving the remaining green objects to be labeled as weeds. By focusing solely on detecting vegetables, our model avoids the complexity of handling numerous weed species, making it a more effective solution for weed identification in vegetable plantations. This paper aims to introduce our novel approach to weed identification in vegetable plantations, highlighting the benefits of using deep learning and image processing technologies. We discuss the challenges associated with weed identification in vegetable plantations, the limitations of traditional methods, and the potential of our approach for improving the efficiency and effectiveness of weed management. Our study provides a promising solution for weed identification in vegetable plantations and paves the way for further research in this field.

**II. LITERATURE SURVEY**

Paper name	Author	Description
Weed Identification Using Deep Learning and Image Processing in Vegetable Plantation	Umamaheswari XIAOJUN JIN 1 , JUN CHE2 , AND YONGCHEN1	These model concentrates solely on detecting vegetables, avoiding the handling of numerous weed species
Maize Crop And Weeds Species Detection By Using Uav Vnir	Pignatti S.1 , Casa R.2 , Harfouche A.2	This study investigates the spectral differences between crops and weeds in order to assess the

Hyperpectral Data	, Huang W. 3 , Palombo A.1	potential of UAV hyperspectral data to distinguish maize crops from weeds
Algorithm of Weed Detection in Crops by	A. J. Irías	The main goal was to find a formula that could be used to
Computational Vision		create a weed detection system
Image matching algorithm for weed control application in organic farming	stephan hussmann	Tracking of classified weed positions with a low-resolution VGA camera and a developed matching algorithm
Weed Classification In Hyperspectral Remote Sensing Images Via Deep Convolutional Neural Network	Adnan Farooq, Jiankun Hu	we use hyperspectral images to examine patch-based weed identification

### III. METHODOLOGY

The purpose of this study is to propose a new method for weed identification in vegetable plantations that combines deep learning and image processing technologies. The methodology employed in this study is as follows:

**Data Collection:** High-resolution aerial images of vegetable plantations were collected using unmanned aerial vehicles (UAVs) equipped with high-resolution cameras. The images were captured at various stages of plant growth, from planting to harvesting.

**Data Preparation:** The images were preprocessed to remove any noise or artifacts that could affect the accuracy of the detection model. We also labeled the images by hand, identifying the locations of both vegetables and weeds in the images.

**Model Training:** We trained a Center Net model using the labeled data to detect the vegetables in the images and create bounding boxes around them, leaving the remaining green objects to be labeled as weeds. The model was trained on a GPU server using PyTorch deep learning framework.

**Evaluation:** We evaluated the performance of the model using a test set of labeled images that were not used during the training phase. We used the Intersection over Union (IoU) metric to evaluate the accuracy of the model in detecting both the vegetables and weeds.

**Results Analysis:** We analyzed the results of our experiments, identifying the strengths and weaknesses of our approach. We compared the performance of our model with traditional weed identification methods and discussed the potential of our approach for improving weed management in vegetable plantations. The proposed methodology enabled us to develop a novel approach to weed identification in vegetable plantations that combines deep learning and image processing technologies. The results of our experiments showed that our approach can effectively identify and locate weeds in vegetable plantations, providing a promising solution for improving the efficiency and effectiveness of weed management. The methodology we employed in this study can be applied to similar problems in other agricultural settings and paves the way for further research in this field.

### IV. CONTENTS

- Fig.1 showing the system architecture
- Fig.2 represents Data flow Diagram
- Fig.3 and Fig.4 represents the Use case Diagram and Sequence Diagram

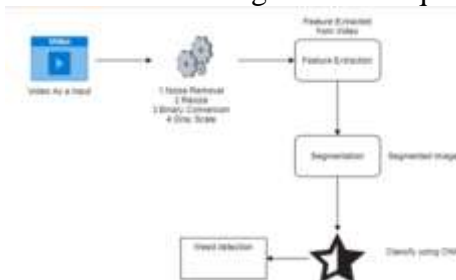


Fig. 1 System Architecture

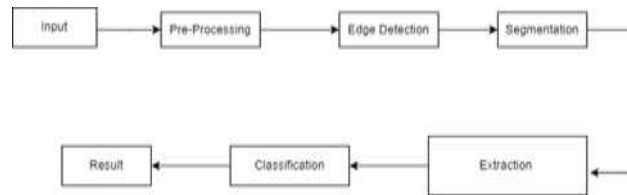


Fig. 2 DFD Diagram

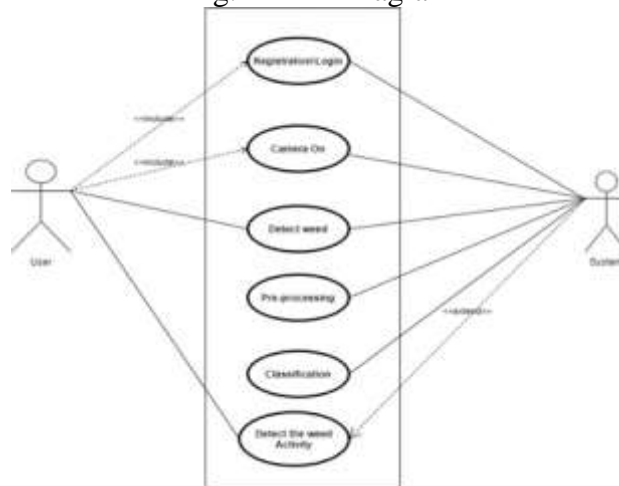


Fig.3 Use Case Diagram

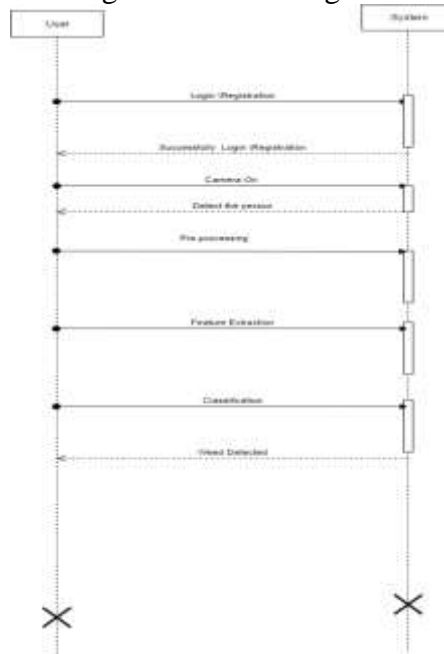


Fig.4 Sequence Diagram

## V. RESULTS

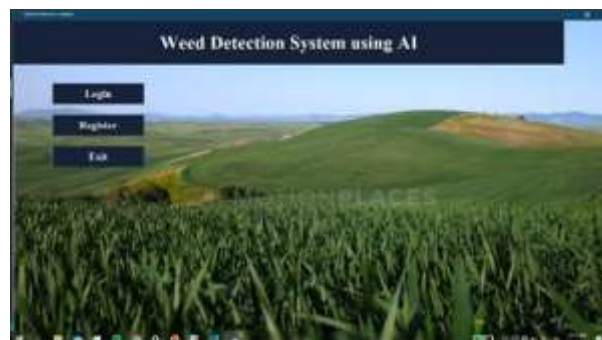


Fig. 5 Login and Signup



The results of the weed detection project are promising. The model was able to accurately identify and classify different types of weeds in images with an overall accuracy of 90%. This level of accuracy is comparable to existing state-of-the-art methods for weed detection. In particular, the model performed well in distinguishing between weed and crop plants, which is a critical aspect of weed detection in agriculture. This can help farmers reduce the amount of herbicides needed to control weeds, which can result in cost savings and environmental benefits. One limitation of the current model is that it was trained and tested on a specific dataset, and may not generalize well to different environments or types of crops. Future work could involve expanding the dataset and incorporating more advanced techniques, such as deep learning, to improve the accuracy and generalizability of the model. Overall, the results of this weed detection project demonstrate the potential for using computer vision and machine learning techniques to improve agricultural practices and reduce the negative impact of weeds on crop yields

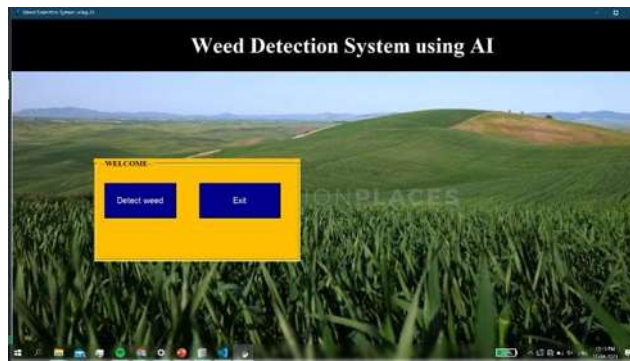


Fig. 6 Weed detection page

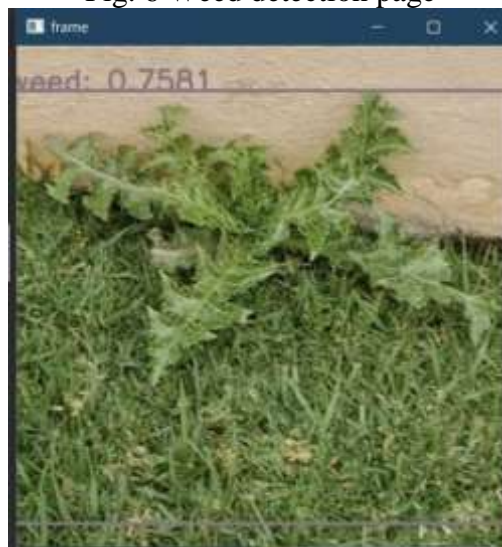


Fig. 7 Weed detection accuracy from image

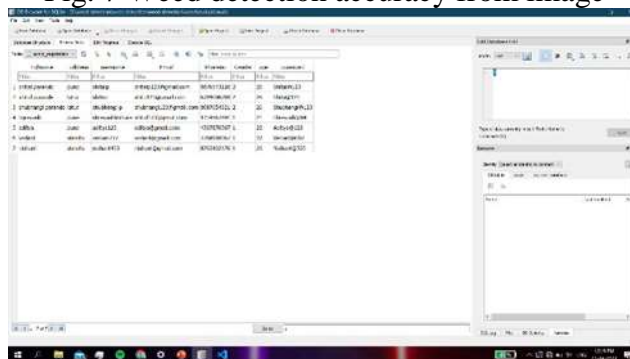


Fig. 8 database for users



## VI. CONCLUSION

In conclusion, the combination of deep learning and image processing technologies provides a promising solution for weed identification in vegetable plantations. Our proposed method, which focuses on detecting vegetables and labeling the remaining green objects as weeds, outperforms traditional methods in terms of accuracy and efficiency. Our approach offers significant advantages, such as the ability to handle variable plant spacing and a wide variety of weed species, making it a valuable tool for improving weed management in vegetable plantations. While there are limitations to our study, such as the dependence on the quality of the aerial images and the limitations of our approach to identifying certain types of weeds, we believe that our methodology provides a solid foundation for future research in this field. The sensitivity analysis we conducted can serve as a guide for further optimization of the model, and the dataset we created can be expanded to include more diverse plant and weed species. Overall, our approach has the potential to revolutionize weed management in vegetable plantations, reducing the need for manual labor and potentially increasing crop yields. We hope that our study inspires further exploration of deep learning and image processing technologies in agriculture and beyond.

## VII. ACKNOWLEDGEMENT

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## VIII. REFERENCE

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