



# PLANT DISEASE DETECTION USING CONVOLUTIONAL NEURAL NETWORK

**Prof. Ranjana Thakuria**

Assistant Professor  
Dept of CSE, SVCE  
Bangalore, Karnataka

**Jeevitha N**

Department of ISE  
SVCE, VTU University  
Bangalore, Karnataka

**Lavanya S**

Department of ISE  
SVCE, VTU University  
Bangalore, Karnataka

**Kruthika S**

Department of CSE  
SVCE, VTU University  
Bangalore, Karnataka

**Sahana M G**

Department of CSE  
SVCE, VTU University  
Bangalore, Karnataka

## ABSTRACT

This study investigates the use of Convolutional Neural Networks (CNNs) for plant disease detection, a critical aspect of ensuring global food security. Through the analysis of recent advancements, the paper outlines the methodology involving dataset collection, preprocessing, and CNN model training. Various CNN architectures and transfer learning techniques are explored for improved classification performance. Strategies such as data augmentation are discussed to enhance model robustness. Experimental results demonstrate the efficacy of CNN-based approaches in achieving high accuracy and reliability in disease detection across diverse crop species. Despite challenges like dataset imbalance and overfitting, CNNs show significant promise in revolutionizing agriculture by enabling timely disease diagnosis, thereby bolstering crop productivity and sustainability.

**Keywords:** Convolutional Neural Network, Dataset.

## I. INTRODUCTION

Ensuring crop health is crucial for global food security, yet plant diseases pose significant challenges, leading to yield losses. Timely disease detection is vital for effective management. Advancements in artificial intelligence, particularly Convolutional Neural Networks (CNNs), offer promising solutions. CNNs excel in image classification, making them ideal for plant disease detection, revolutionizing agricultural practices.

This paper explores CNN utilization for plant disease detection. Our methodology involves training CNN models on large datasets of healthy and diseased plant leaves. Through convolutional layers, activation functions like ReLU, and max-pooling layers, these models extract features indicative of diseases. Integration of dense layers and dropout regularization enhances prediction accuracy.

Additionally, we delve into supplement utilization for plant health management. Supplements play a crucial role in enhancing plant resilience and overall vigor. Incorporating supplement information into disease detection provides valuable insights for farmers.

This paper aims to provide an overview of CNN-based plant disease detection and supplement usage. By integrating technology with practical agricultural strategies, we contribute to sustainable agriculture and global food security.

## II. LITERATURE SURVEY

"Identification of Various Diseases in Plant Leaves Using Image Processing and CNN Approach" by Diksha Tandekar, Snehlata Dongre, DOI: 10.1109/ICCCNT56998.2023.10306979, 23 November 2023. This study employs deep learning techniques, specifically VGG Net 19 and CNN, to detect diseases in plant leaves. With accuracies of 85.4% and 83% respectively, these models offer promising results. Moreover, the proposed approach not only identifies diseases but also suggests remedies, enhancing the overall management of crop health." "An Ensemble-Based Model of Detecting Plant Disease using CNN and Random Forest" by Shalya Saxena; Sandeep Rathor, DOI: 10.1109/ISCON57294.2023.10112023, 04 May 2023. This proposed framework achieves 99.89 % accuracy during the testing/validation phase on 380000 images taken from standard datasets. Also, the plant disease is detected using RF and CNN separately and then, the accuracy of RF, CNN, and CNN with RF models are compared.

"Ladies Finger Leaf Disease Detection using CNN" by R Renugadevi; S Vaishnavi; S. Santhi; S Pooja, DOI: 10.1109/ICAAIC56838.2023.10140976, 04-06 May 2023. This paper provides a comprehensive literature review of research efforts in plant disease detection, focusing on both traditional hand-crafted feature-based methods and advanced deep learning approaches. While hand-crafted features have been utilized for recognizing diseases in ladies finger plants, the adoption of deep learning techniques addresses limitations inherent in manual feature extraction. Deep learning models demonstrate superior performance compared to manually defined criteria, showcasing their effectiveness in plant disease diagnosis.

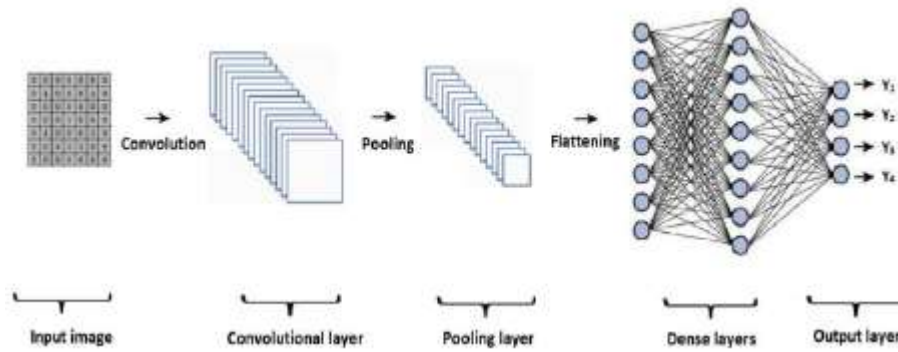
"A Detection and Classification of Cotton Leaf Disease Using a Lightweight CNN Architecture" by Arvind.K. S; Archana Negi, DOI: 10.1109/ICERECT56837.2022.10060246, 15 March 2023. In this paper, it is proposed to train a deep learning Faster R-CNN model using the cotton crop leaf dataset in order to identify and classify leaf diseases. The Plant Village dataset, along with VGG-16, InceptionV1, and V2, is used as a benchmark in the process of determining which feature extractor is the most effective.

"Real-Time Plant Leaf Disease Detection using CNN and Solutions to Cure with Android App" by Rohit Chandra Joshi; Vivek Raj Patel; Abhijeet Mishra; Sandeep Kumar, DOI: 10.1109/ICCCIS60361.2023.10425034, 15 February 2024. In the current digital era, mobile phones are ubiquitous, presenting an opportunity to enhance accessibility to plant disease information. This paper discusses an interactive mobile app utilizing a Convolutional Neural Network (CNN) for real-time plant disease identification and treatment guidance. Users can simply upload images of diseased plants to receive immediate disease diagnosis and access treatment recommendations. The paper outlines the application's structure, along with the data and results generated by the plant disease detection model, facilitating effective disease detection and treatment.

"EnC-SVMWEL: Ensemble Approach using CNN and SVM Weighted Average Ensemble Learning for Sugarcane Leaf Disease Detection" by Uvarani Vignesh; Bala Subramanian Chokkalingam, DOI: 10.1109/ICSCDS56580.2023.10104818, 25 April 2023. This paper introduces a novel approach, EnC-SVMWEL, for sugarcane plant leaf disease detection. Utilizing an ensemble of Convolutional-Support Vector Machine Weighted Average Learning, sugarcane leaf images are processed through grayscale conversion, resizing, and contrast enhancement. Feature extraction is conducted using DenseNet201 architecture, and the Support Vector Machine Weighted Average Learning classifier is employed for classification into six disease classes. Evaluation metrics including accuracy, precision, recall, and F-measure demonstrate the effectiveness of the proposed approach, achieving a remarkable detection accuracy of 97.45%, surpassing existing methods.

"Detection of Plant Disease for Paddy Crop Using an Ensemble of CNNs", DOI: 10.1109/ICRTAC59277.2023.10480787, 02 April 2024. This research introduces a novel method for detecting diseases in paddy crops, aiming to enhance agricultural productivity and quality. By employing a weighted voting mechanism to combine multiple CNN architectures, the proposed method improves disease categorization precision and robustness. Extensive testing on a large dataset demonstrates that the ensemble CNNs outperform individual CNNs and current best practices. Notably, the ensemble technique enhances both accuracy and stability in diagnosing diseases, even amidst variations in image quality and symptoms. This study contributes to precision agricultural operations and sustainable crop management by efficiently utilizing CNN ensembles. The findings underscore the importance of employing deep learning techniques to enhance current methods of disease detection and prevention, crucial for ensuring the safety of the world's food supply.

### III. SYSTEM ARCHITECTURE AND METHODOLOGY



**Figure 1 – System Architecture**

Methodology for Plant Disease Detection Using CNN:

#### 1. Data Collection and Preparation:

- Gathered a diverse dataset of plant images, including healthy specimens and those afflicted by various diseases, sourced from public repositories.
- Conducted data preprocessing, which involved resizing images to a standard 224x224 pixel size and normalizing pixel values to [0, 1].

#### 2. Model Architecture and Training:

- Employed a CNN model architecture comprising convolutional layers for feature extraction, ReLU activation functions for non-linearity, max-pooling layers for downsampling, and fully connected layers for classification.
- Trained the model using batch-wise processing with the Adam optimizer and categorical cross-entropy loss function, iterating over multiple epochs to adjust model weights.

#### 3. Hyperparameter Tuning and Optimization:

- Conducted hyperparameter tuning to optimize model performance, adjusting parameters such as learning rate, batch size, dropout rate, and kernel sizes using techniques like grid search and random search.

#### 4. Deployment and Integration:

- Developed a user-friendly web application using Flask, HTML, CSS, and JavaScript to provide an intuitive interface for interacting with the plant disease detection system.
- Integrated the trained CNN model into the backend of the web application, deploying it on a cloud platform for scalability and reliability.
- Managed user data using a database management system and implemented continuous integration and deployment pipelines for rapid updates.

#### 5. User Interaction:

- Designed the user interface of the web application with a focus on simplicity, clarity, and ease of use, incorporating intuitive navigation and interactive elements.
- Implemented user-friendly image upload features with real-time feedback during image processing and prediction.
- Presented prediction outcomes, disease information, prevention steps, and supplement recommendations in a clear and structured format to enhance comprehension.

### IV. CONCLUSION

In conclusion, the utilization of Convolutional Neural Networks (CNNs) for plant disease detection presents a promising solution to address the challenges faced by the agricultural sector in mitigating the impact of plant diseases on crop yields and global food security. Through the implementation of a robust methodology encompassing data collection, model training, hyperparameter tuning, and deployment, CNN-based models have demonstrated remarkable capabilities in accurately identifying diseased plant specimens from images.

The successful deployment of CNN models in plant disease detection systems has enabled farmers and agricultural stakeholders to benefit from timely and precise disease diagnosis, facilitating proactive disease management strategies

and the implementation of targeted interventions. By harnessing the power of deep learning algorithms, these systems have contributed to increased crop productivity, reduced economic losses, and enhanced agricultural sustainability.

In summary, plant disease detection using CNNs represents a pivotal advancement in agricultural technology, offering transformative solutions to safeguard crop health, ensure food security, and support the sustainable development of global agriculture.

## V. REFERENCES

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