



## Identification of Nutritional Deficiency and Disease Prediction of Crops using Deep Learning: A Review

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health and yield by giving farmers real-time, actionable insights, which eventually helps to boost agricultural sustainability and efficiency.

### ABSTRACT-

The rapid and accurate identification of crop illnesses and malnutrition is essential to ensuring sustainable agricultural practices and maximizing crop yields. This project presents a novel approach to diagnosing crop health issues by leveraging Convolutional Neural Networks (CNNs), in particular, are deep learning methods for the automated detection of nutrient deficiencies and diseases from leaf images. The system uses advanced image processing algorithms to analyze leaf patterns, textures, and color variations, identifying specific symptoms associated with common plant nutrient deficiencies, such as nitrogen, phosphorus, and potassium, as well as diseases like leaf blight and mildew. To guarantee excellent prediction accuracy and real-time applicability, the model is trained on a variety of datasets of tagged crop leaf images. This technology enables prompt interventions to enhance crop health and yield by giving farmers real-time, actionable insights, which eventually helps to boost agricultural sustainability and efficiency.

**Keywords-** Nutritional Deficiency, Deep Learning, Crop Disease Prediction, Agricultural Automation-

### 1. Introduction -

Ensuring optimal crop health is crucial for maximizing agricultural productivity, but identifying nutrient deficiencies and diseases can be challenging using traditional manual methods, which are often slow and prone to error. With advancements in deep learning and image processing, these challenges can be addressed through automation.

This project uses Convolutional Neural Networks (CNNs) to analyze leaf images, detecting crop nutrient deficiencies and diseases in real-time. It enables timely interventions, improving crop health and promoting sustainable farming, while overcoming the inefficiencies of traditional manual diagnosis.

The objective of this project is to create a system based on deep learning that accurately identifies and classifies nutrient deficiencies and diseases in crops through automated image analysis.

### 1.1 Block Diagram

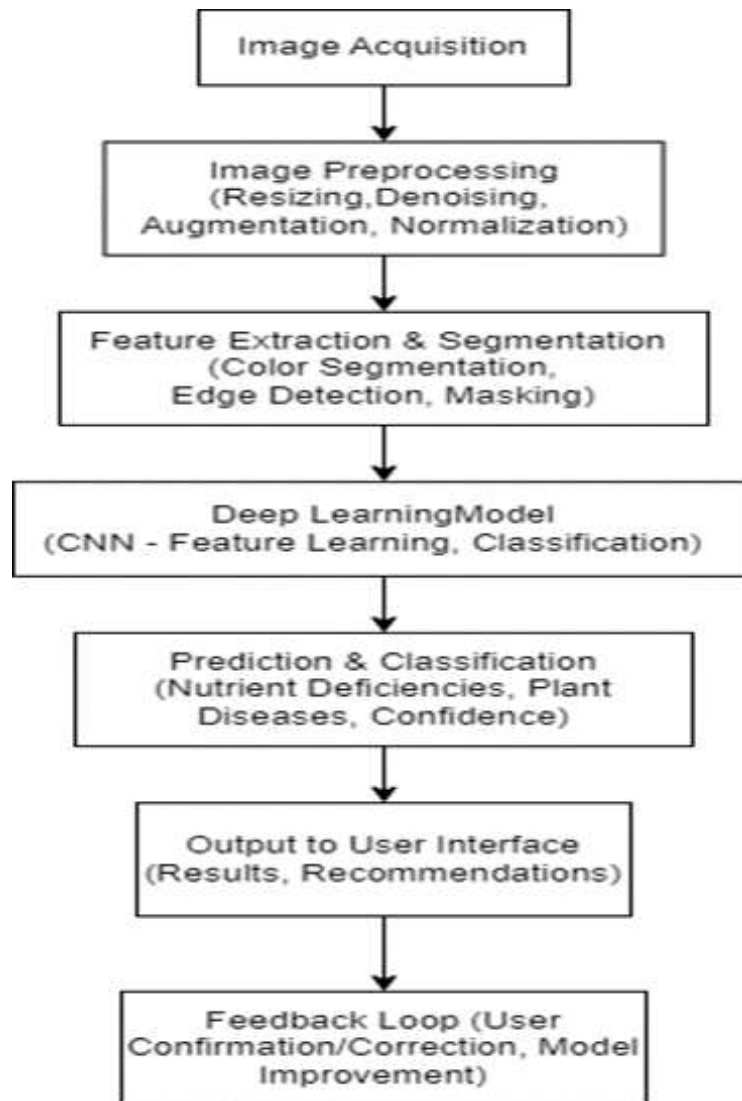


Fig. Block Diagram of Identification of Nutritional Deficiency and Disease Prediction System

In this project, farmers or users capture images of crop leaves using smartphones, drones, or field cameras, which are then preprocessed by resizing, denoising, augmenting, and normalizing to ensure consistent input quality. Key features such as leaf color, spots, and edges are extracted, while irrelevant areas are masked. A convolutional neural network (CNN) processes these features to categorize nutrient deficiencies or diseases, providing predictions along with confidence scores. The results, along with treatment recommendations, are displayed in a user-friendly interface. A feedback loop allows users to confirm or correct predictions, continuously improving the model's accuracy over time.

## 2. Literature Survey:

This paper reviews Identification of Nutritional Deficiency and Disease prediction of Crops using Deep Learning, emphasizing CNNs, with many implementations achieving over 75% accuracy.

[1] This work reviews computer vision-based technologies for early detection of crop nutrient deficiencies to improve crop productivity and health. [2] The study examines the application of machine learning and proximate imaging for detecting plant diseases and nutrient deficiencies, highlighting progress, challenges, and future research directions in digital farming. [3] This research uses deep learning techniques, particularly CNN-SVM, to accurately classify nutrient deficiencies in maize plants with 98% accuracy, enhancing crop productivity and sustainability [4] A Multi- Attention Convolutional Neural Network effectively identifies plant nutritional deficiencies with high accuracy using attention mechanisms and spatial pooling, enhancing crop management

[5] This project develops an Android application using machine learning and image processing to automatically identify nutrient shortages and pest infestations in coconut leaves, achieving over 93% accuracy. [6] An AI-based system using deep learning models, particularly DenseNet, is developed to

diagnose tomato plant diseases with 95.31% accuracy in real-time. [7] A deep learning model using a Convolutional Neural Network (CNN) detects NPK nutrient deficiencies in rice plants with 98.75% accuracy. An app for Android has been created to help farmers to access fertilizer recommendations and improve crop yield.

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### 3. Identification of Nutritional Deficiency and Disease Prediction of Crops using Deep Learning

The system aims to supply farmers with feasible real-time insights to enhance crop health and yield. By integrating advanced image processing and CNN models, it streamlines diagnosis and offers tailored recommendations for interventions. Ultimately, the project seeks to enhance agricultural productivity, sustainability, and resource efficiency.

#### 3.1 System Architecture

Here's a concise architecture for Nutritional Deficiency and Disease Prediction in Crops using Deep Learning:

**Image Acquisition:** High-quality images of symptomatic crop leaves are captured using smartphones, drones, or sourced from agricultural databases. These images are uploaded to a centralized system for analysis.

**Preprocessing:** Images are standardized through resizing, denoising, augmentation, and normalization to enhance quality and facilitate model training. This step ensures efficient processing and better model performance.

**Feature Extraction and Segmentation:** Focus on symptomatic areas of leaves using techniques like color segmentation, edge detection, and masking to highlight relevant features. This targets the specific regions affected by deficiencies or diseases.

**Deep Learning (CNN):** Identify vitamin inadequacies in photos using Convolutional Neural Networks (CNNs). (e.g., nitrogen, phosphorus) or diseases (e.g., blight, mildew). Transfer Learning or Ensemble Models can enhance performance.

**Prediction and Classification:** Multi-class classification for both deficiencies and diseases with confidence scores.

**Output:** Display results via a user-friendly app, providing advice on nutrient management or disease treatment.

**Feedback Loop:** Gather user feedback to continuously improve model accuracy. This system integrates deep learning to detect crop issues and provide actionable insights to farm.

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### 4. Benefits of Identification of Nutritional Deficiency and Disease Prediction on Crops

#### 4.1 Enhanced Accuracy

- CNNs provide significantly greater accuracy in identifying and classifying nutrient deficiencies and diseases in crops compared to traditional methods.
- By training on large datasets, CNNs improve diagnostic precision, even for visually subtle or ambiguous symptoms.

#### 4.2 Automation

- The automation of crop image analysis using CNNs and image processing techniques eliminates time-consuming manual inspections, enabling efficient monitoring of large-scale farms.
- This faster diagnosis facilitates timely interventions, helping to prevent crop damage and maintain optimal growth conditions.

#### 4.3 Real-Time Feedback

- The system provides real-time analysis of crop images, offering farmers instant feedback on plant health, enabling them to quickly assess issues.
- This enables farmers to take prompt corrective actions, such as adjusting fertilizers or treatments, to prevent disease spread and mitigate nutrient deficiencies before substantial crop loss occurs.

### 5. Main Objectives

Here are the main objectives for the project:

**Automation of Diagnostics:** To automate the image processing and analysis workflow, reducing the need for manual inspections and accelerating the diagnostic process for large-scale agricultural operations.

**Real-Time Insights:** To provide farmers with real-time feedback on crop health, enabling them to make timely and informed decisions regarding nutrient management and disease treatment.

**User-Friendly Interface:** To design an intuitive user interface that displays analysis results and recommendations in a clear and actionable format, enhancing user experience for farmers.

**Continuous Improvement:** To establish a feedback loop that allows the system to learn from user input and improve its accuracy and effectiveness over time, adapting to changing agricultural conditions.

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## 6. Conclusion

In conclusion, this project leverages deep learning and image processing techniques to precisely detect and categorize crop illnesses and nutrient shortages. By automating the diagnostic process and providing real-time insights, it empowers farmers to make timely interventions, reducing crop loss and improving yield. The system's integration of predictive analytics further enhances its capability, enabling proactive management of crop health. With a user-friendly interface and a continuous feedback loop for improvement, this solution promises to streamline agricultural workflows and contribute to more sustainable farming practices.

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