



# AgriSmartHub: Empowering farmers with ML and NLP

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**Abstract :** Agriculture is a cornerstone of India's economy, employing a vast portion of the population. To meet increasing food demands, it is essential to enhance farming practices. Advanced technologies like machine learning (ML) and deep learning (DL) offer the potential to significantly boost agricultural productivity through early crop disease detection and optimized resource use. Our project, AgriSmartHub, utilizes ML techniques such as KNN, SVM, and DL methods like CNN to detect plant diseases swiftly and accurately. By training these models on extensive datasets, they effectively identify patterns and predict diseases. Specifically, our DL system automates the scanning of leaf images to identify diseases based on visual symptoms, assesses disease severity, and recommends suitable fertilizer quantities. A user-friendly interface allows farmers to easily capture leaf images and receive insightful, tailored suggestions, aiding in improved crop production and quality. AgriSmartHub empowers farmers to make informed decisions and optimize yields.

**Keywords** – AgriSmartHub, nlp, ml, convolutional neural network, artificial intelligence, crop, disease, fertilizers, machine learning, natural language processing, crop recommendation, Agriculture, farming.

## 1. INTRODUCTION

In India, agriculture stands as the backbone of the economy, supporting a significant portion of the workforce and playing a pivotal role in sustaining livelihoods. However, the sector faces numerous challenges, including the need to meet rising food demands, optimize resource usage, and combat crop diseases effectively. In response to these challenges, leveraging advanced technologies such as machine learning (ML) and natural language processing (NLP) has emerged as a promising solution to revolutionize agricultural practices.

The AgriSmartHub project aims to address these pressing issues by harnessing the power of ML and NLP to empower farmers with actionable insights and recommendations for enhanced farming outcomes. By integrating cutting-edge techniques such as k-nearest neighbors (KNN), support vector machines (SVM), and deep learning (DL), particularly convolutional neural networks (CNN), our project endeavors to swiftly and accurately detect plant diseases, thus mitigating crop losses and improving overall agricultural productivity.

At the core of our project lies the development of robust ML models trained on extensive datasets of crop images. These models are designed to analyze visual symptoms and patterns associated with various crop diseases, enabling early detection and diagnosis. Specifically, our DL system automates the process of scanning leaf images, distinguishing between healthy and diseased plants, and assessing the severity of the detected diseases. Moreover, it provides tailored recommendations for appropriate fertilizer usage, thereby assisting farmers in mitigating the impact of diseases on crop health and optimizing yields.

One of the key strengths of the AgriSmartHub project is its user-friendly interface, designed to cater to the diverse needs

of farmers and agricultural workers. Through a simple leaf image capture process, users can access insightful suggestions and recommendations tailored to their specific crop conditions. This intuitive interface not only facilitates ease of use but also empowers farmers to make informed decisions in real-time, leading to improved crop production and quality.

By bridging the gap between advanced technologies and agricultural practices, AgriSmartHub aims to transform the way farming is approached in India. Through the integration of ML and NLP, our project seeks to empower farmers with the knowledge and tools necessary to enhance productivity, optimize resource utilization, and ensure sustainable agricultural development.

In the following sections of this research paper, we delve deeper into the methodologies employed, the results obtained, and the implications of the AgriSmartHub project for the agricultural sector in India and beyond.

## 2. LITERATURE SURVEY:

The identification of crop diseases and fertilizers using machine learning techniques and algorithms has gained significant attention in recent years due to its potential to streamline and improve the agriculture process. Several studies and research papers have explored various aspects of crop disease prediction and fertilizer recommendation in different contexts. Here is a literature survey highlighting some key research in this field:

1. "Crop Recommendation System using Machine Learning" by Dhruvi Gosai, Chintal Raval et al. (2021): This paper presents a comprehensive framework for improving crop production and soil health through the synergistic integration of IoT and ML technologies. The proposed system offers practical solutions to address the challenges faced by farmers and agricultural stakeholders.

2. "Fertilizers Recommendation System for Disease Prediction Using Machine Learning" by Sindu. P, R. Shoba Rani, G. Victo Sudha George, J. Jayaprakash et al. (2024): The authors propose a machine learning-based recommendation system to increase agricultural productivity. In this work, sophisticated models were devised and developed to estimate crop yield, recommend fertilizer, and identify plant sickness.

3. "Harvestify - Crop Disease Detection and Fertilizer Suggestion using CNN" by Sachin Adulkar, Vivek Pawar, Aniket Choudhari, Shubham Kothekar, Shruti Agrawal et al. (2019): This research paper explores a solution for agriculture that can assist farmers in increasing their overall productivity by monitoring the agricultural field. Rainfall reserves and soil boundaries. The study highlights the importance of NLP in enhancing the efficiency of agriculture.

### 3. Requirement:

The Entire Development Process Has Been Subdivided Into Two: the Front End Development and the Backend Development. The Front End Comprises of the Visually Visible Parts Such as the Main Page, User Login Page, User Sign in Page and Home page. The Back End Contains the Database and Its Interaction With the Front-end.

#### 1) Front End Development:

For front end HTML, CSS, JavaScript and Bootstrap is used. Our aim is to develop a system with responsive design and is user friendly. The design developed is simple and clean to understand by any user. HTML defined a structure of the website and for styling CSS is used. To add the functionality in the website JavaScript is used. Bootstrap enhanced the development process by minimizing the time required to develop a design.

#### 2) Backend Development:

Backend is developed using JavaScript and Python. Machine learning and NLP is used for image processing and give the crop disease predictions. Along with this fertilizer recommendations, crop recommendation is provided to the farmers on successful analysis of soil reports. GPU resources are utilized for faster training, and the LIME technique is applied to interpret model predictions.

### 4. RESEARCH METHODOLOGY:

This paper presents a comprehensive overview of the AgriSmartHub application, which integrates IoT and machine learning technologies to optimize crop production and soil health. The application comprises various modules, each designed to address specific agricultural challenges, including fertilizer recommendation, disease detection, crop suggestions, and news feed implementation. The following sections provide detailed insights into the design, implementation, and evaluation of each module:

#### A. The Application

**1. Recommendation for Fertilizer:** The module enables users to receive personalized fertilizer recommendations based on soil parameters and crop type. Users input nitrogen, phosphorus, and potassium values along with the crop name, triggering a POST request to the Flask API. The API, hosted at a specified location, processes the input data and provides fertilizer recommendations. The front-end receives an HTTP response, displaying the recommended fertilizer to the user.

**2. Disease Detection:** This module facilitates the early detection of plant diseases by analyzing leaf images. Users can either upload images directly or capture them using the application. The images are processed by the backend, and the model predicts the presence of diseases. Upon processing, an HTTP response is sent to the front-end, presenting the identified diseases and corresponding treatment options.

**3. Recommended Crop:** Users can obtain recommendations for suitable crops based on soil parameters such as nitrogen, phosphorus, and potassium levels. Similar to the fertilizer recommendation module, a POST request is made to the Flask API, triggering model inference. The API responds with recommendations for the best crop yield based on the input soil parameters.

**4. Disease Website:** The application includes a disease portal that offers comprehensive information on various plant diseases and recommended products for treatment. Users can access this portal to gain insights into disease management strategies.

**5. Evaluation of Interpretability:** This module employs the LIME technique to provide interpretability to model predictions. The user's plant leaf image is sent to a deployed API, where LIME computation is performed. The resulting image, along with interpretive insights, is displayed on the front-end, aiding users in understanding the model's decision-making process.

#### B. A Crop Machine Learning Proposal: The Set's Description

This section outlines the datasets and machine learning models utilized in the application. The dataset for crop prediction consists of seven key features, including soil nutrient levels, temperature, humidity, and pH values. Six models, including Decision Tree, Naive Bayes, SVM, and XGBoost, are evaluated using cross-validation on the dataset. Similarly, for disease recognition, the PlantVillage dataset is employed, containing RGB images of healthy and diseased crops. Three pre-trained models, VGG-16, ResNet-50, and EfficientNetB0, are utilized for image classification tasks. The models are trained using Adam optimization with categorical cross-entropy loss and early stopping mechanisms. GPU resources are utilized for faster training, and the LIME technique is applied to interpret model predictions, enhancing model transparency and user understanding. To select the most effective fertilizer for a plant, we employed rule-based classification, a classification scheme that makes use of IF-THEN rules for class prediction. A fertilizer may be required depending on how far a plant is from its ideal N, P, or K value. For our motivations, we have 6 sorts of manure suggestions right now, in view of whether the N/P/K qualities are high or low.

Overall, the AgriSmartHub application presents a holistic solution for enhancing agricultural practices through the integration of IoT and machine learning technologies. By providing personalized recommendations for fertilizer usage, early detection of crop diseases, and suggestions for crop selection, the application empowers farmers to make informed decisions, thereby improving crop productivity and soil health. The use of interpretability techniques further enhances user trust and understanding of model predictions, fostering adoption and acceptance in the agricultural community.

### 5. RESULTS AND DISCUSSION:

The AgriSmartHub platform efficiently parses and analyzes agricultural data, extracting vital information such as crop health, soil quality, and farming practices. This data is organized into structured formats, facilitating detailed analysis and decision-making for farmers and agricultural stakeholders. With the goal of optimizing crop production and enhancing agricultural practices, the platform offers insights into trends, disease patterns, and optimal resource utilization. Real-time data analysis enables quick access to relevant information, empowering users to make informed decisions and improve overall agricultural productivity. AgriSmartHub serves as a valuable tool for farmers, researchers, and policymakers, driving innovation and sustainability in the agricultural sector.

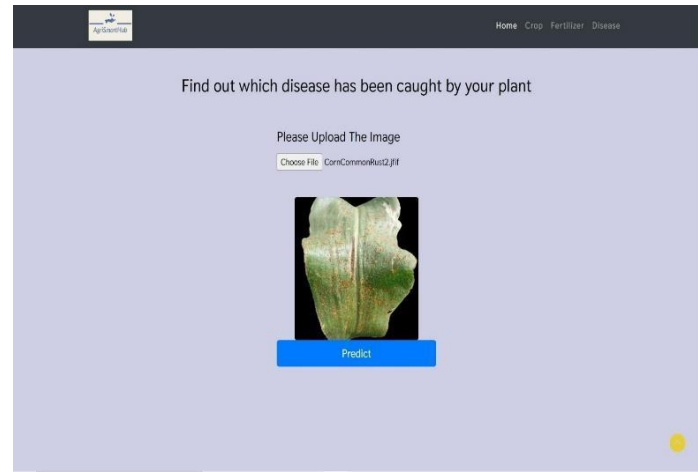


Fig.5.3 Upload crop image

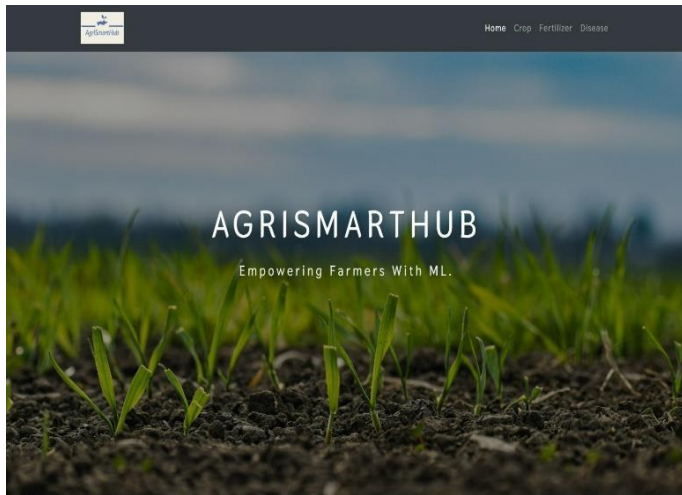


Fig.5.1. Home Page

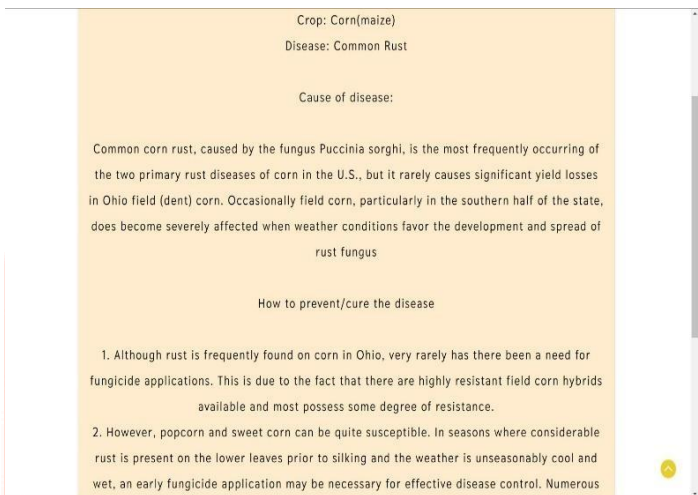


Fig.5.4 Output

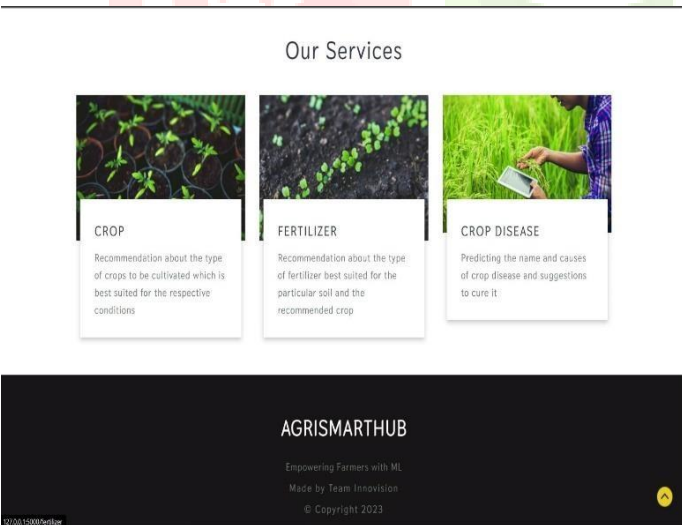


Fig.5.2. Home Page services section

The AgriSmartHub application efficiently processes agricultural data, extracting critical insights like crop health and soil quality. By storing information in a structured format, it enables swift analysis for optimizing farming practices. The system's goals include trend identification and resource utilization, enhancing agricultural productivity. With streamlined data retrieval, it empowers users to make informed decisions, driving innovation and sustainability in farming.

## Discussion:

AgriSmartHub emerges as a transformative tool in the agricultural sector, offering streamlined solutions for farmers and agricultural stakeholders. By integrating machine learning and IoT technologies, it facilitates efficient analysis and management of agricultural data. The platform's ability to provide personalized recommendations for fertilizer usage, crop selection, and disease detection marks a significant advancement in agricultural practices. Moreover, its user-friendly interface and automated processes simplify complex tasks, empowering farmers to make informed decisions and optimize yields. AgriSmartHub stands as a beacon of innovation, revolutionizing farming practices and contributing to sustainable agricultural development.

## 6. Diagram:

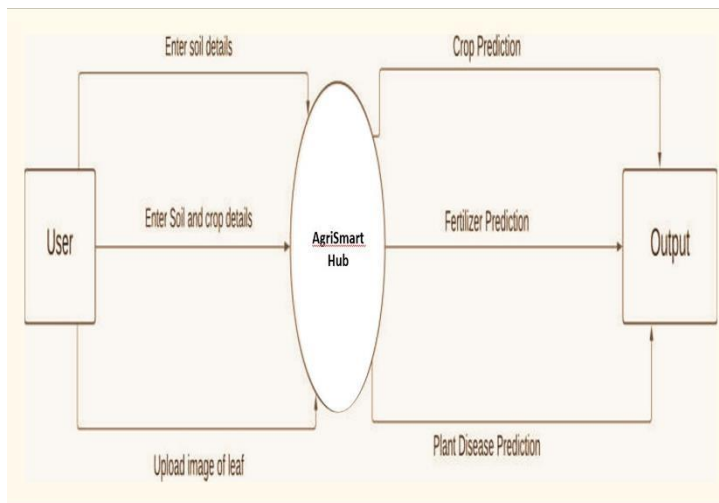


Fig.6.1. Flow Diagram

## 7. Conclusion:

In the ever-evolving landscape of agriculture, optimizing farming practices and empowering farmers with advanced tools are paramount for sustainable food production. The AgriSmartHub project represents a pioneering solution that harnesses the capabilities of Machine Learning (ML) and Natural Language Processing (NLP) to revolutionize agricultural productivity. With its user-friendly interface and advanced features, including crop disease detection, fertilizer recommendations, and crop selection suggestions, AgriSmartHub empowers farmers to make informed decisions and maximize yields. By bridging the gap between technology and agriculture, this project heralds a new era of innovation and efficiency in farming practices, ensuring food security and economic prosperity in India and beyond.

## 8. Acknowledgment:

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