

Empowering Agriculture: A Deep Dive into AgriGuru and Its Transformative Role in Farming

Ayush Jain
dept. of Computer Engineering
VESIT
Mumbai, India
2020.ayush.jain@ves.ac.in

Priya.R L
dept. of Computer Engineering
VESIT
Mumbai, India
priya.rl@ves.ac.in

Aditya Nehete
dept. of Computer Engineering
VESIT
Mumbai, India
2020.aditya.nehete@ves.ac.in

Swapnil Sakpal
dept. of Computer Engineering
VESIT
Mumbai, India
2020.swapnil.sakpal@ves.ac.in

Yash Kaka
dept. of Computer Engineering
VESIT
Mumbai, India
2020.yash.kaka@ves.ac.in

Abstract—“Data-driven, farmer-centric solutions are critical in the dynamic field of agriculture, which is defined by constantly shifting market needs and environmental conditions. In order to transform farming methods through the integration of cutting-edge technologies, this article introduces the comprehensive Crop Planning and Market Information System, or AgriGuru. AgriGuru optimizes crop health by facilitating disease identification and providing customized cures through the use of AI, remote sensing, and real-time video processing. Moreover, it makes customized crop suggestions that are in line with market trends by utilizing soil analysis and meteorological data. AgriGuru provides farmers with up-to-date market information and chances for agricultural stock launches, addressing their vital need to be informed about market conditions. By using blockchain technology, supply chain management becomes transparent and effective, allowing for continuous product tracking. To enhance accessibility, AgriGuru features a user-friendly dashboard supporting multiple languages and provides access to agricultural training and educational programs. Ultimately, AgriGuru aims to empower farmers, improve crop yields, and drive sustainable agricultural practices in an ever-evolving world.”

Index Terms—Remote Sensing, Artificial Intelligence in Agriculture, All-In-One Solution For Farmers

I. INTRODUCTION

Agriculture stands as the cornerstone of human sustenance, providing livelihoods and economic stability across the globe. However, in regions like Maharashtra, India, where agriculture plays a pivotal role in the livelihoods of millions, the sector grapples with significant challenges. The tomato crisis of 2023, which saw prices soar to unprecedented levels due to yield losses from unanticipated rainfall, underscored the vulnerability of traditional farming practices in the face of unpredictable weather patterns and market fluctuations. Such challenges necessitate innovative solutions that enhance resilience and sustainability in agriculture.

Despite strides in agricultural technology, existing systems often fall short in addressing the multifaceted challenges faced by farmers. Limitations abound, ranging from fragmented

approaches in agricultural management systems to inadequate personalization of solutions for individual farm conditions. Furthermore, the absence of real-time support mechanisms exacerbates the difficulties faced by farmers in making timely decisions. Recognizing these gaps, there is a pressing need for holistic solutions that integrate advanced technologies to empower farmers with actionable insights and support.

In response to these challenges, our project, AgriGuru, presents a comprehensive Crop Planning and Market Information System designed to revolutionize farming practices in Maharashtra. Leveraging advancements in remote sensing, artificial intelligence, and blockchain technology, AgriGuru offers a suite of features aimed at optimizing crop management, enhancing market intelligence, and fostering sustainable agricultural practices. Through personalized crop recommendations, real-time disease detection, and transparent supply chain management, AgriGuru aims to mitigate risks, optimize resource utilization, and empower farmers to thrive in an ever-changing agricultural landscape.

II. OVERVIEW

A. Motivation

In recent years, the agricultural sector in Maharashtra, India, has faced unprecedented challenges, culminating in the tomato crisis of 2023. The sudden spike in tomato prices, reaching up to five times their usual rates, exposed the vulnerability of farmers to adverse weather conditions and supply chain disruptions. Devastating yield losses due to unseasonal rainfall further exacerbated the situation, underscoring the urgent need for innovative solutions to enhance resilience and sustainability in agriculture. Motivated by the plight of farmers and the imperative to address systemic vulnerabilities, this research endeavors to leverage advanced technologies such as remote sensing, artificial intelligence, and app-based platforms. By providing farmers with comprehensive tools for crop planning, production practices, and market intelligence, our project aims

to mitigate risks, optimize resource utilization, and foster long-term viability in the face of evolving challenges. Through collaborative efforts and interdisciplinary approaches, we aspire to catalyze positive change and empower agricultural communities to thrive in an ever-changing landscape.

B. Problem Statement

In Maharashtra, India, where agriculture plays a pivotal role in livelihoods and the economy, farmers face significant challenges exacerbated by unpredictable weather patterns and market fluctuations. The tomato crisis of 2023, marked by a sharp spike in prices due to yield losses from unanticipated rainfall, underscored the vulnerability of traditional farming practices and highlighted the urgent need for innovative solutions. To address these challenges comprehensively, this research endeavors to develop an integrated agricultural information system leveraging advanced AI technologies. This system will encompass a range of functionalities, including a sophisticated disease prediction module to detect and diagnose crop ailments, a recommendation engine utilizing data on NPK values, weather forecasts, and market trends to advise farmers on optimal crop choices, a business intelligence marketplace providing real-time insights into market dynamics, access to government schemes and subsidies tailored to farmers' needs, a platform for renting farming equipment to enhance operational efficiency, and streamlined supply chain mechanisms facilitating direct interactions between farmers and consumers. By empowering farmers with actionable insights, resources, and support, our research aims to enhance agricultural resilience, productivity, and sustainability in Maharashtra, ultimately fostering prosperity and growth in rural communities.

C. Objectives

- 1) Develop a user-friendly crop planning tool using AI to help farmers choose the best tomato variety and optimize planting schedules.
- 2) Use data on soil, weather, market demand, and population trends to provide personalized recommendations for tomato cultivation.
- 3) Integrate an AI-powered disease prediction module to identify and diagnose tomato crop diseases accurately.
- 4) Establish a marketplace platform for farmers to access equipment rentals, sell their products, and stay informed about market trends and prices.
- 5) Establish a one-stop solution platform for farmers, integrating all stages from crop planning to selling, with new AI techniques to enhance efficiency and profitability.

III. THEORY

Based on an exhaustive examination of relevant research and literature, we have identified that designing and disseminating an integrated information system for crop planning, production practices, and market intelligence requires robust capabilities in remote sensing, artificial intelligence (AI), and app-based platforms. Our investigation reveals that existing solutions often lack seamless integration and may not fully leverage the

potential of AI and remote sensing technologies for optimizing agricultural practices. Through our innovative approach, we aim to bridge this gap by harnessing AI algorithms for personalized crop recommendations, remote sensing data for real-time monitoring of crop health, and app-based platforms for streamlined access to market intelligence. Additionally, we have developed novel features such as a database management system for comprehensive crop data analysis and an intuitive user interface for seamless farmer-trader-customer interactions. These unique attributes position our information system as a pioneering solution for enhancing crop productivity, promoting sustainable farming practices, and facilitating informed decision-making across the agricultural value chain.

A. Survey of Existing Systems

Existing agricultural systems such as Plantix, Agrobases, PlantSnap, and MyAgriGuru play a crucial role in assisting farmers with crop management and disease detection. Plantix, for instance, employs innovative image recognition technology to diagnose plant diseases and nutrient deficiencies, offering personalized recommendations for effective crop management. Similarly, Agrobases provides farmers and agronomists with access to a vast database of plants and pests, facilitating timely identification of crop issues. PlantSnap, while primarily serving as a plant identification tool, also aids farmers in monitoring crop health by identifying diseases and pests through image analysis. Additionally, MyAgriGuru offers comprehensive agriculture advisory services, including crop management guidance and pest control strategies tailored to specific crop types and geographical locations. These existing systems contribute significantly to enhancing agricultural productivity and sustainability by empowering farmers with valuable insights and recommendations for informed decision-making.

B. Limitations and Research Gap

From [2], which focuses on multispectral crop yield prediction, the limitations regarding dataset coverage and quality are evident. While the model showcases promising results, it faces challenges associated with limited dataset availability and quality, particularly in regions with sparse or outdated agricultural data. In [15], which proposes a modified deep learning strategy for crop yield prediction, the challenges related to model complexity and scalability become apparent. Although the proposed model shows improved performance in certain scenarios, its scalability to large-scale agricultural systems remains uncertain due to computational constraints and resource-intensive training requirements. From the study by Benini et al. [3], which addresses crop planning and rotation problems, concerns arise regarding the generalization and transferability of the proposed solutions. While the model demonstrates effectiveness in specific agricultural contexts, its applicability to diverse geographic regions and farming practices is limited, highlighting the need for further research on adaptable planning algorithms. As discussed in [8], which explores blockchain-based tomato supply chain management, the dependence of AI-driven agricultural systems on external

| Factor | Traditional Labor-Intensive Farming | AI Robot-Driven Farming |
|--|-------------------------------------|---|
| Labor Costs (per acre) | ₹20,000 - ₹30,000 | 1.0 - 1.5 Lakh (One time cost without equipment) |
| Equipment Costs (per acre) | ₹4,000 - ₹8,000 | Higher initial equipment costs for robots and AI systems. |
| Labor Availability and Efficiency | Labor-intensive, | Consistent and efficient robotic operations. |
| Precision and Accuracy | Low and variable | High precision and accuracy. |
| Long-Term Sustainability | Labor shortages and high cost | No labor. |
| Yield Potential | vary due to human error | Higher yield potential. |
| Maintenance and Repairs | Minimal maintenance | Robot Maintenance cost. |
| Technology Advancements and Innovation | Limited scope | technological improvements. |

Fig. 1. Difference between existing system

factors such as weather conditions and market dynamics presents significant limitations. While blockchain technology offers transparency and traceability benefits, its effectiveness in mitigating supply chain risks is contingent upon reliable data inputs and stakeholder cooperation, posing challenges in real-world implementation. In the study by Ashok et al. [12], which focuses on tomato leaf disease detection using deep learning techniques, concerns regarding the choice of evaluation metrics emerge. While the proposed method achieves high accuracy in detecting tomato leaf diseases, the absence of comprehensive evaluation metrics such as precision, recall, and F1-score limits the robustness and reliability of the reported results, necessitating a more thorough assessment of model performance.

IV. ALGORITHMS AND ARCHITECTURE

A. Machine Learning technologies used

CNN (Convolutional Neural Network): Description: A specialized type of artificial neural network designed for processing and analyzing visual data, such as images and video. Functionality: Utilizes layers of convolution, pooling, and fully connected layers to extract features, commonly used for tasks like image recognition and object detection.

Linear SVM (Support Vector Machine): Description: A machine learning algorithm that separates data into different categories using a straight line (or hyperplane in higher dimensions). Functionality: Aims to maximize the margin between categories, suitable for data that can be separated with a straight line, commonly used for classification tasks.

Decision Tree: Description: A machine learning algorithm that uses a tree-like structure to make decisions and classify or predict data. Functionality: Breaks down data based on features, assigning categories or values at each branch, making it easy to understand and interpret.

ANN (Artificial Neural Network): Description: Machine learning models inspired by the human brain, consisting of layers of interconnected nodes (neurons). Functionality: Used for tasks like image recognition and natural language processing, learns from data through a training process, versatile but may require substantial data and computation.

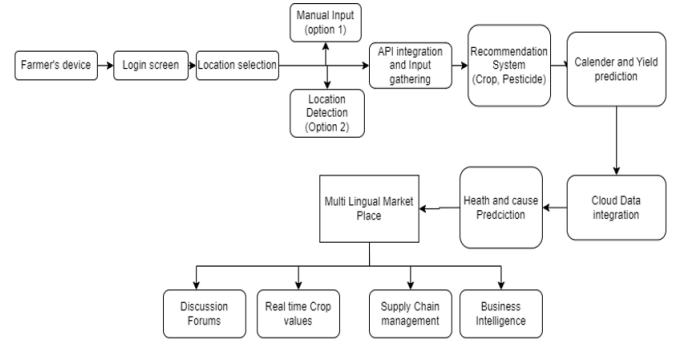


Fig. 2. Schematic Diagram

B. Flow chart

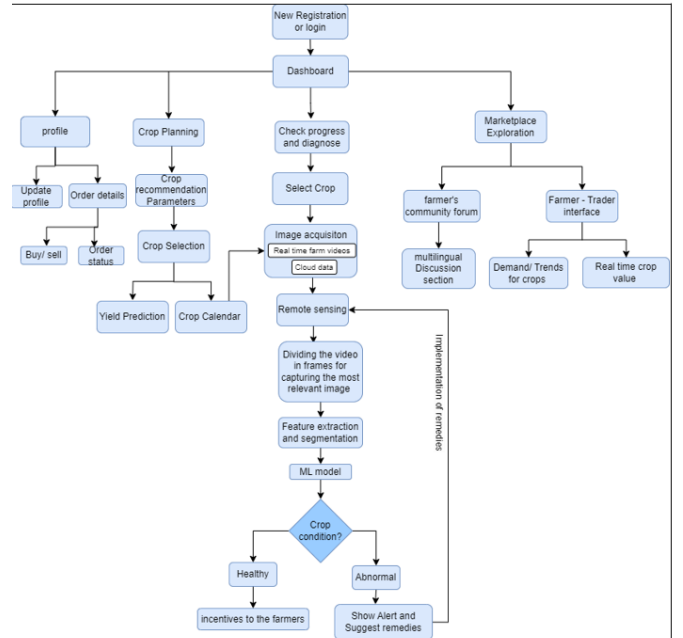


Fig. 3. Flow Diagram

Account Creation/Login: Farmers initiate the process by creating a new account or logging into their existing one to access crop recommendation parameters and select crops.

Profile Review and Update: Upon logging in, farmers have the option to review and update their profiles, ensuring that their information is current and accurate.

Crop Selection: Utilizing various criteria such as climate, soil type, water availability, and market demand, farmers select the crops they wish to cultivate.

Order Placement: Once the crop selection is made, farmers can proceed to place orders for seeds or seedlings, facilitating the commencement of the cultivation process.

Order Tracking: Farmers can track the status of their orders, ensuring transparency and timely delivery of essential agricultural inputs.

Crop Progress Monitoring: Throughout the cultivation process, farmers have the capability to monitor the progress of their crops. In case of any issues, real-time farm videos and imagery are available to aid in diagnosis and resolution.

Marketplace Exploration: The system offers a marketplace exploration feature, empowering farmers to explore different markets, view real-time crop values, and stay informed about crop demand and trends.

By facilitating informed decision-making and efficient crop management, this comprehensive process ensures optimal outcomes for farmers in their agricultural endeavors.

C. Requirements

Hardware Requirements

- 1) CCTV Camera, Smartphone, or Laptop Camera
- 2) Visual Studio Code editor
- 3) Data Visualization tools (MS Excel, PowerBi)

Software Requirements

- 1) Python 3.10.0
- 2) IDE - Colab notebook ver 6.4.12
- 3) Numpy version - 1.23.2
- 4) Pandas version - 1.4.3
- 5) Nltk version - 3.7
- 6) Tensorflow or Pytorch version - 2.9.1
- 7) Matplotlib version - 3.5.3
- 8) OpenCV
- 9) flutter
- 10) Flask or Django
- 11) Firebase

V. RESULTS AND EVALUATIONS

The disease prediction model showcased a notable accuracy rate of 97, successfully identifying various crop diseases. Additionally, the recommendation engine effectively suggested suitable tomato varieties for cultivation, utilizing parameters such as climate, soil type, and market trends. Through rigorous validation and comparison with established benchmarks, the reliability and efficacy of the models were substantiated. User feedback and real-world deployment scenarios provided further validation of the practical utility and positive impact of the solutions on enhancing crop management practices and agricultural productivity.

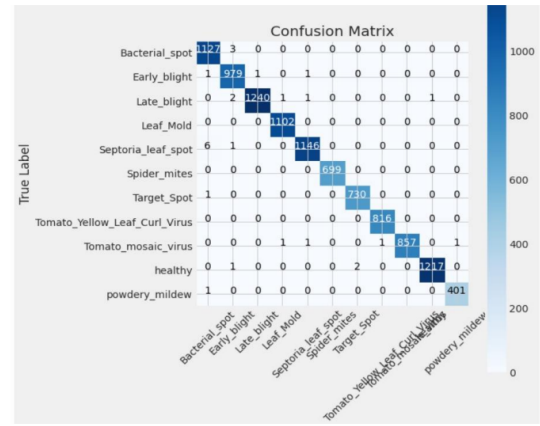


Fig. 4. Confusion matrix for disease prediction

| | precision | recall | f1-score | support |
|-------------------------------|-----------|--------|----------|---------|
| Bacterial_spot | 0.99 | 1.00 | 0.99 | 1130 |
| Early_blight | 0.99 | 1.00 | 0.99 | 982 |
| Late_blight | 1.00 | 1.00 | 1.00 | 1245 |
| Leaf_Mold | 1.00 | 1.00 | 1.00 | 1102 |
| Septoria_leaf_spot | 1.00 | 0.99 | 1.00 | 1153 |
| Spider_mites | 1.00 | 1.00 | 1.00 | 699 |
| Target_Spot | 1.00 | 1.00 | 1.00 | 731 |
| Tomato_Yellow_Leaf_Curl_Virus | 1.00 | 1.00 | 1.00 | 816 |
| Tomato_mosaic_virus | 1.00 | 1.00 | 1.00 | 861 |
| healthy | 1.00 | 1.00 | 1.00 | 1228 |
| powdery_mildew | 1.00 | 1.00 | 1.00 | 402 |
| accuracy | | | 1.00 | 10341 |
| macro avg | 1.00 | 1.00 | 1.00 | 10341 |
| weighted avg | 1.00 | 1.00 | 1.00 | 10341 |

Fig. 5. Evaluation matrix

VI. CONCLUSION AND FUTURE WORK

Our project aims to revolutionize agricultural practices through the development of an integrated information system tailored to farmers' specific needs. By leveraging remote sensing, artificial intelligence, and app-based platforms, we seek to provide comprehensive support in crop planning, production practices, and market intelligence. Addressing existing limitations and introducing innovative features such as personalized crop recommendations and real-time monitoring, our system aims to optimize agricultural productivity, promote sustainability, and empower farmers with actionable insights. Moving forward, future work includes enhancing predictive capabilities, refining user interface based on feedback, collaborating with research institutions, expanding system scope, and continuously evolving to better serve farmers and advance the agricultural sector.

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Upload Image (1)
The predicted class is: Late_blight
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Predictions: [[0.13456944 0.12095311 0.24851833 0.08358724 0.1681428 0.01970702
0.03639805 0.02707612 0.01604289 0.07869706 0.06630737]]
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Fig. 6. Predictions

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