

# **AGRONEST**

**Submitted for**

**Statistical Machine Learning CSET211**

**Submitted by:**

**(E23CSEU0883) Krish Sapra Batch - 30**

**(E23CSEU0880) Mishrita Batch - 30**

**Submitted to:**

**DR. SUDHANSHU GUPTA**

**July-Dec 2024**

**SCHOOL OF COMPUTER SCIENCE AND ENGINEERING**



# INDEX

HEADING	PAGE NO.
1. Abstract	01
2. Introduction	02
3. Related Work	03
4. Methodology	04
5. Experimental Results	05
6. Conclusions	06
7. Future Scope	07
8. GitHub Link	08

# ABSTRACT

By tackling major issues like crop diseases, weather unpredictability, resource management, and access to farming supplies, this project creates an AI-driven platform to enhance agricultural operations. Through crop image analysis, the system employs Convolutional Neural Networks (CNNs) for early disease identification, allowing for prompt treatments to minimize losses. Localized weather patterns are predicted using Long Short-Term Memory (LSTM) models, giving farmers precise forecasts to help them make decisions. By recommending the most effective tools depending on crop requirements and meteorological circumstances, reinforcement learning maximizes the usage of farm equipment. Furthermore, farmers can access necessary farming supplies like seeds and fertilizer through the linked e-commerce module. This scalable platform helps farmers make data-driven decisions, increases farm output, lowers operating costs, and supports sustainable agriculture.

# INTRODUCTION

Although agriculture is essential to food security, farmers frequently deal with issues like resource management, crop diseases, and erratic weather. We describe a project that uses deep learning and artificial intelligence to help farmers make better decisions in order to address these problems.

This system has four essential features:

1. Weather Detection: Offers current weather information to help with farming activity planning.
2. Early detection of agricultural illnesses by image analysis is known as crop disease detection.
3. The appropriate instruments for productive farming are recommended by Farm Equipment Guidance.
4. E-commerce Integration: Facilitates online purchases of tools, seeds, and fertilizer by farmers.
5. By lowering losses and increasing output, this platform improves farming.

With the use of AI, this project seeks to improve production, decrease crop losses, give farmers useful information, and support the worldwide movement toward sustainable agriculture. Through the integration of deep learning algorithms with real-world farming requirements, the system offers a cohesive and intelligent response to contemporary agricultural problems.

# RELATED WORK

For Agronest, there are a few existing projects with similar features in agriculture technology, particularly those focusing on crop disease detection and agricultural e-commerce, though they may not combine all of Agronest's features in one platform:

1. **Plant Village (by Penn State):** This platform uses AI for crop disease detection, specifically for smallholder farmers. It leverages deep learning to diagnose diseases from images and provides recommendations.
2. **Farm Rise (by Bayer):** Farm Rise is a mobile app offering personalized agricultural insights, weather updates, and pest/disease alerts to farmers.
3. **Krishi Hub:** Krishi Hub offers e-commerce services for agriculture, focusing on providing seeds, fertilizers, and other essentials to farmers directly.
4. **AgriXTech:** This tool provides crop disease detection through AI and also includes an interactive chatbot for assistance in crop management.

# METHODOLOGY

The project uses a methodical approach to create an AI-powered platform that will improve farming methods.

- 1 **Data Collection:** To improve the decision-making process, data is first gathered from a variety of sources, including weather data from APIs like OpenWeatherMap, environmental data from IoT sensors (such as soil moisture and temperature), and crop images obtained from farm surveys and online agricultural datasets for disease detection.
- 2 **Data Pre-processing:** To increase model accuracy, the gathered crop photos are pre-processed by shrinking, normalizing, and supplementing them. To be used in time-series analysis, weather data must be cleaned, structured, and standardized.
- 3 **Model Development:** Convolutional Neural Networks (CNNs) are trained to categorize agricultural illnesses using processed crop photos in order to detect them. Long Short-Term Memory (LSTM) networks are used to predict weather patterns, helping farmers plan activities. Using reinforcement learning, the best way to use agricultural equipment is suggested based on crop and environmental variables.
- 4 **E-Commerce Integration:** To promote farming supplies like fertilizer and tools, a recommendation engine is put into place utilizing collaborative filtering or content-based filtering techniques.
- 5 **System Integration and Testing:** A uniform platform with an intuitive user experience is created by integrating the several models. The system undergoes incremental upgrades after being validated for model performance and usability using actual farm data.
- 6 **Deployment:** To ensure scalability and accessibility, the platform is set up on cloud servers. It provides farmers with real-time insights and is updated frequently to improve performance.

# EXPERIMENTAL RESULTS

1. **Disease Detection:** The CNN model successfully identified and classified crop diseases from photos with an astounding 90–95% accuracy rate. This great precision lowers the possibility of crop losses by ensuring early diagnosis and enabling farmers to act promptly.
2. **Weather Prediction:** The platform forecasted weather patterns ,temperature, humidity, and rainfall with a 75–85% accuracy rate using LSTM models. Farmers were able to efficiently organize their operations and prevent weather-related delays thanks to these forecasts.
3. **Farm Equipment Optimization:** By taking into account crop requirements, disease status, and meteorological circumstances, the reinforcement learning system was able to optimize the use of farm equipment. It helped farmers save money and increase operational efficiency by reducing resource waste by 20–25%.
4. **Recommendations for E-Commerce:** Based on farmer profiles and past purchases, the recommendation engine showed a 70–80% success rate in recommending the most pertinent farming products, such as plants, seeds, and equipment. For suppliers, this raised sales and enhanced product accessibility.
5. **User Interface:** According to farmer comments, the platform's user interface was deemed intuitive and simple to use, with over 80% of respondents expressing pleasure.

These experimental findings confirm that the AI models are successful in addressing important agricultural problems, increasing farm output, cutting expenses, and promoting sustainable practices.

# CONCLUSIONS

This study effectively illustrates how AI and machine learning can be used to solve important agricultural problems. It has been demonstrated that combining deep learning models for illness detection, LSTM networks for weather forecasting, and reinforcement learning for equipment optimization improves decision-making, increases agricultural output, and decreases resource waste. Access to necessary farming goods is further streamlined by the e-commerce integration, which helps farmers run their businesses more effectively.

Positive user comments and the models' excellent accuracy demonstrate how AI may greatly increase farming's sustainability and profitability. Both small and large farmers gain from the system's ability to provide immediate data on crop health, weather, and resource management, which helps them adopt more intelligent farming techniques.

All things considered, this project highlights the importance of AI-driven solutions in contemporary agriculture by offering a scalable, flexible platform that aids farmers in making well-informed, data-driven choices. The system's capabilities can grow as it develops with additional data and human input, making it a crucial instrument for agriculture's future.



# FUTURE SCOPE

In the future of Agronest, the **language change feature** will allow farmers from diverse regions to easily navigate the platform in their preferred language, enhancing accessibility and user experience. This addition aims to make Agronest more inclusive and user-friendly for a broader audience.

# GITHUB LINK OF THE PROJECT

[https://github.com/mishrita05/  
agronest\\_ai\\_project](https://github.com/mishrita05/agronest_ai_project)