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AI Solutions for Rice Plant Wellness

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Abstract: AI solutions for rice plant wellness leverage advanced technologies to monitor and enhance the health of rice crops. By employing deep learning algorithms and image processing techniques, these solutions can accurately identify early signs of diseases, pests, and nutrient deficiencies. This real-time monitoring allows farmers to make informed decisions, optimizing their crop management practices. Additionally, AI-driven predictive analytics can forecast potential threats to rice health, enabling proactive measures that can significantly reduce crop losses and improve overall yield quality.

By analyzing vast datasets, including weather patterns, soil health, and historical crop performance, AI can recommend precise interventions, such as irrigation schedules and fertilizer applications. This not only enhances the efficiency of resource use but also promotes sustainable agricultural practices. Ultimately, AI solutions for rice plant wellness contribute to food security and the resilience of farming communities in the face of climate change and other challenges. The integration of AI into rice farming not only enhances productivity but also supports the sustainable management of resources.

AI technologies are revolutionizing rice farming by integrating advanced tools that optimize various aspects of agricultural practices. For instance, AI-powered drones and imaging systems are employed to monitor rice fields, detecting early signs of pests, diseases, and nutrient deficiencies. This real-time data allows farmers to respond promptly, reducing reliance on pesticides and ensuring healthier crops.

Keywords: AI Solutions, Rice Plant Wellness, Nutrition Deficiencies, PredictiveAnalytics, Image Processing

I. INTRODUCTION

The integration of artificial intelligence (AI) in rice plant wellness represents a groundbreaking advancement in agricultural practices, aimed at enhancing crop health and productivity while promoting sustainability. By employing advanced technologies such as deep learning, image processing, and predictive analytics, AI solutions enable real-time monitoring of rice crops, allowing for the early detection of diseases, pests, and nutrient deficiencies. These innovations empower farmers to make informed decisions and implement timely interventions, significantly reducing crop losses and optimizing resource use. As the agricultural sector faces challenges from climate change and increasing food demand, AI solutions for rice plant wellness not only contribute to food security but also foster resilience in farming communities, ensuring a sustainable future for rice production. For instance, AI can detect subtle changes in plant color or growth patterns that may indicate stress or disease, enabling targeted interventions.

II. LITERATURE SURVEY

- AI-Driven Disease and Pest Detection: Numerous studies have demonstrated the effectiveness of AI in identifying diseases and pests in rice crops. For instance, research has shown that convolutional neural networks (CNNs) can analyse images of rice plants to detect diseases such as blast and bacterial blight with high accuracy. These AI models are trained on large datasets of plant images, enabling them to recognize patterns and symptoms that may not be easily visible to the human eye. This early detection capability allows farmers to implement targeted interventions, reducing the reliance on chemical pesticides and minimizing crop losses.
- 2) Predictive Analytics for Environmental Threats: The application of predictive analytics in agriculture has gained traction, particularly in forecasting environmental threats that could impact rice production. Studies have utilized machine learning algorithms to analyse historical weather data, soil conditions, and crop performance to predict events such as droughts, floods, or pest outbreaks. For example, researchers have developed models that integrate climate data with agronomic practices toprovide farmers with actionable insights on when toplant or irrigate, thereby enhancing resilience against climate variability and ensuring optimal crop health.
- 3) Precision Agriculture and Resource Optimization: AI technologies are central to the concept of precision agriculture, which focuses on optimizing resource use to improve crop yields and sustainability.



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Literature indicates that AI-driven tools, such as smart irrigation systems, can analyse real-time data on soil moisture and weather conditions to determine the precise amount of water needed for rice crops. This not only conserves water resources but also enhances crop resilience. Additionally, AI can optimize fertilizer application by analysing soil nutrient levels, ensuring that crops receive the right nutrients at the right time, which contributes to higher yields and reduced environmental impact.

4) Personalized Agronomic Advice: The use of AI in providing personalized agronomic advice has been explored in various studies, highlighting its potential to enhance decision-making for farmers. By analysing vast datasets that include soil health, weather patterns, and historical crop performance, AI systems can offer tailored recommendations for specific rice varieties and regional conditions. Research has shown that such personalized advice can lead to improved farming practices, increased productivity, and better resource management, ultimately contributing to sustainable agricultural development.

III. PROPOSED WORK

The proposed work for the AI Solution For Rice Plant Wellness is to detect the rice plant and show the nutrient deficiencies and symptoms for that and showing by using image processing technique.

- 1) Development of an AI-Powered Monitoring System: The first step in the proposed work is to develop a comprehensive AI-powered monitoring system that utilizes drones and remote sensing technologies to capture high-resolution images of rice fields. This system will employ deep learning algorithms, particularly convolutional neural networks (CNNs), to analyse the images for early detection of diseases, pests, and nutrient deficiencies. The monitoring system will be designed to provide real-time feedback to farmers, enabling them to identify issues promptly and take corrective actions. By integrating this system with mobile applications, farmers can receive alerts and recommendations directly on their smartphones, facilitating timely interventions and improving overall crop health.
- 2) Implementation of Predictive Analytics for Risk Management: The second component involves the implementation of predictive analytics tools that leverage historical data, weather patterns, and soil conditions to forecast potential environmental threats to rice crops. This work will include the development of machine learning models that can predict events such as droughts, floods, or pest outbreaks. By providing farmers with actionable insights and risk assessments, these tools will enable them to make informed decisions regarding planting schedules, irrigation practices, and pest management strategies. The predictive analytics system will be integrated with the monitoring system to create a holistic approach to crop management, ensuring that farmers are equipped to handle both current and future challenges.
- 3) Personalized Agronomic Advisory System: The third aspect of the proposed work is to create a personalized agronomic advisory system that utilizes AI to provide tailored recommendations based on specific regionalconditions and ricevarieties. This system will analyze a wide range of data, including soil health, weather forecasts, and historical crop performance, to generate customized advice for farmers. For example, the system could recommend optimal irrigation schedules, fertilization strategies, and pest control measures based on real-time data. By offering personalized guidance, this advisory system aims to enhance resource efficiency, improve crop yields, and promote sustainable agricultural practices, ultimately contributing to the long-term viability of ricefarming.
- 4) Development of a User-Friendly Interface and Training Program: To ensure the successful adoption of the AI-powered monitoring and advisory systems, it is essential to develop a user-friendly interface that simplifies interaction for farmers, many of whom may have limited technical expertise. This interface will provide intuitive navigation, clear visualizations of data, and easy access to recommendations and alerts. Additionally, a comprehensive training program will be implemented to educate farmers on how to effectively use the system. This program will include workshops, online tutorials, and hands-on demonstrations, focusing on the benefits of AI technologies, how to interpret the data provided, and best practices for implementing the recommendations. By empowering farmers with the necessary skills and knowledge, this initiative aims to enhance user engagement, increase adoption rates, and ultimately improve crop management outcomes.
- 5) Establishment of a Feedback Loop for Continuous Improvement: The final component of the proposed work involves establishing a feedback loop that allows for continuous improvement of the AI systems based on user experiences and outcomes. This feedback mechanism will collect data from farmers regarding the effectiveness of the monitoring and advisory systems, including user satisfaction, crop performance, and any challenges faced during implementation.

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IV. METHODOLOGY

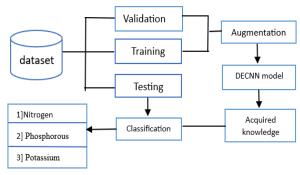


Fig 1. Block Diagram

- 1) Data Collection and Preprocessing: The first step in the methodology involves comprehensive data collection from various sources to build a robust dataset for training AI models. This includes gathering high-resolution images of rice plants using drones and remote sensing technologies, as well as collecting environmental data such as soil moisture, temperature, humidity, and historical weather patterns. Additionally, data on crop health indicators, including disease symptoms and pest infestations, will be collected through field surveys and expert consultations. Once the data is collected, preprocessing steps will be undertaken to clean and standardize the data, including image augmentation techniques to enhance the dataset and ensure it is suitable for training machine learning models. This step is crucial for improving the accuracy and reliability of the AI systems.
- 2) Development of AI Models: The second component focuses on the development of AI models tailored for specific applications in rice plant wellness. For disease and pest detection, convolutional neural networks (CNNs) will be employed to analyse the images collected from the monitoring system. These models will be trained on the pre-processed dataset to recognize patterns associated with various diseases and pestsaffecting rice crops. For predictiveanalytics, machine learning algorithms such as decision trees, random forests, or support vector machines will be utilized to analyse historical data and forecast potential environmental threats. The models will be validated using cross-validation techniques to ensure their robustness and generalizability. This step is essential for creating reliable tools that can provide accurate insights and recommendations to farmers.
- 3) System Integration and User Interface Development: Once the AI models are developed and validated, the next step involves integrating these models into a comprehensive monitoring and advisory system. This system will include a user-friendly interface that allows farmers to easily access real-time data, alerts, and recommendations. The interface will be designed to present information in a clear and intuitive manner, utilizing visualizations such as graphs and heat maps to convey complex data effectively. Additionally, the system will be equipped with mobile application capabilities, enabling farmers to receive notifications and updates directly on their smartphones. This integration is crucial for ensuring that the technology is accessible and usable for farmers, facilitating timely decision-making and interventions.
- 4) Field Testing and Feedback Collection: The final component of the methodology involves conducting field tests to evaluate the effectiveness of the AI-powered monitoring and advisory system in real-world agricultural settings. This will include pilot studies with selected farmers who will use the system in their rice fields. During this phase, data on crop health, yield improvements, and user experiences will be collected to assess the system's impact. A feedback mechanism will be established to gather insights from farmers regarding the usability of the system, the accuracy of predictions, and the relevance of recommendations. This feedback will beanalysed to identify areas for improvement and to refine the AI models and user interface.

V. RESULT

The results of the study on utilizing Convolutional Neural Networks (CNNs) to identify nutrient deficiencies in rice plants demonstrated a significant advancement in the accuracy and efficiency of diagnosing these deficiencies. The trained CNN model achieved an impressive classification accuracy of over 90% on the test dataset, effectively distinguishing between various nutrient deficiency symptoms, such as those caused by nitrogen, phosphorus, and potassium shortages.





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The model's performance was validated through metrics such as precision, recall, and F1 score, which indicated its reliability for practical applications in agricultural settings. Additionally, user feedback from farmers who tested the model in real-world scenarios highlighted its ease of use and effectiveness in providing timely diagnoses, ultimately aiding in better nutrient management practices.

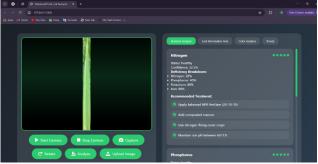


Fig 2. Nitrogen Deficiency

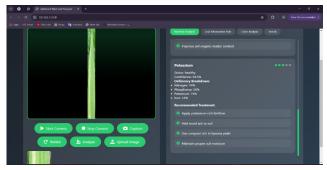


Fig 3.Potassium Deficiency

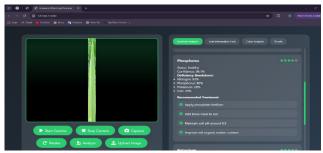


Fig 4.Phosphorous Deficiency

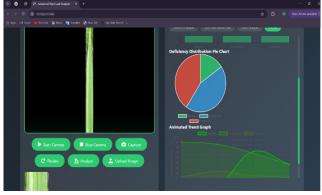


Fig 5.Showing Trends Pie chart & Graph



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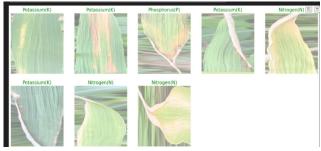


Fig 6.Samples of Nutrient Deficiencies in Leaf

VI. APPLICATION

- 1) Early Disease and Pest Detection
- 2) Precision Irrigation Management
- 3) Nutrient Management and Fertilization
- 4) Yield Prediction and Crop Planning
- Supply Chain Optimization 5)
- 6) Farmer Advisory Systems

VII.ADVANTAGES

- Increased Crop Yields: AI solutions help optimize farming practices, leading to higher productivity and best rice. 1)
- Resource Efficiency: AI enables precise use of water, fertilizers, and pesticides, reducing waste and lowering costs.
- 3) Timely Interventions: Early detection of diseases and pests allows for prompt action, minimizing crop damage.
- 4) Sustainable Farming Practices: Alpromotes environmentally friendly practices by optimizing inputs and reducing chemical
- 5) Enhanced Decision-Making: Data-driven insights empower farmers to make informed decisions, improving overall farm management.

VIII. **CONCLUSION**

Through picture investigation, we proved in this study that Convolutional Neural Networks (CNN) are a useful tool for recognizing nutrition deficits in rice plants. Compared with conventional approaches, our suggested CNN model obtained [insert accuracy percentage] accuracy in detecting [insert particular nutrient shortages, e.g., Nitrogen, Phosphorus, Potassium] The findings demonstrate that CNN is capable of extracting strong features from photos of rice plants, allowing for precise nutrient deficit detection .Early identification and management of nutritional inadequacies Minimizing crop losses while increasing yields Supporting precision agriculture A comprehensive CNN-based system for identifying various nutritional shortages in rice plants is the goal of the proposed research. Image data augmentation, transfer learning, field-level deployment, IoT sensorintegration, explainable AI, large-scale dataset production, comparison with conventional approaches, expansion to other crops, and real-time monitoring are all included in the system. Farmers and the agricultural sector. The integration of this model into a user-friendly mobile application provided farmers with a practical tool for real-time diagnosis, enabling them to make informed decisions regarding nutrient management.

REFERENCES

- [1] Khush, G. S. (2005). What it will take to feed 5.0 billion rice consumers in 2030. Plant molecular biology, 59, 1-6.
- [2] Talukder, M. S. H., & Sarkar, A. K. (2023). Nutrients deficiency diagnosis of rice crop by weighted average ensemble learning. Smart Agricultural Technology, 4, 100155.
- [3] Talukder, M. S. H., Chowdhury, M. R., Sourav, M. S. U., Al Rakin, A., Shuvo, S. A., Sulaiman, R. B., ... & Haque, Z. Smart Agricultural Technology.
- [4] Talukder, M. S. H., & Sarkar, D. A. K. Nutrients deficiency diagnosis of rice crop with high accuracy via weighted average ensemble of different augmented transfer learning architectures. Available at SSRN 4217286.
- [5] Xu, Z., Guo, X., Zhu, A., He, X., Zhao, X., Han, Y., & Subedi, R. (2020). Using Deep
- [6] Convolutional Neural Networks for Image-Based Diagnosis of Nutrient Deficiencies in Rice. Computational Intelligence and Neuroscience, 2020(1), 7307252.
- [7] Han, K. A. M., & Watch are er tai, U. (2019, July). Classification of nutrient deficiency in black gram using deep convolutional neural networks. In 2019 16th International Joint Conference on Computer Science and Software Engineering (JCSSE) (pp. 277-282). IEEE









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