



e-ISSN:2582-7219



INTERNATIONAL JOURNAL OF MULTIDISCIPLINARY RESEARCH IN SCIENCE, ENGINEERING AND TECHNOLOGY

Volume 7, Issue 7, July 2024



INTERNATIONAL
STANDARD
SERIAL
NUMBER
INDIA

Impact Factor: 7.521



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Forecasting the Rice Disease of the Leaves

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ABSTRACT: In the realm of agricultural data, there is a significant need for automated detection and judgement of rice leaf diseases. Machine learning acts a key role in effectively addressing the challenges associated with identifying these diseases. This study introduces a novel machine learning-based approach for disease identification in rice, encompassing several illnesses that frequently impact rice leaves. The methodology involves extracting key features from photos of rice leaf diseases, followed by classification using diverse machine learning techniques. Among these approaches, a quadratic Stand Vector System (SVM) classifier achieved an impressive accuracy of 81.8%. Shape characteristics such as area, roundness, and area-to-lesion ratio were utilized to distinguish between different types of rice illnesses, yielding promising and satisfactory results.

I. INTRODUCTION

Agriculture is a significant part of the Indian economy and is responsible for a second-place contribution to rice production. The majority of Indian states, including Tamil Nadu, West Bengal, Punjab, Uttar Pradesh, Assam, Bihar, etc., cultivate rice. The contribution of the agriculture sector to the overall GDP is roughly 19.9%. In India, Among those is millet. food grains that is most commonly consumed. Diseases have an impact on the quality and growth of rice plants, which therefore affects how profitable farming is. With their limited knowledge from experience, farmers may not has the capacity to identify the different illnesses that could harm a particular harvest of rice. In food production, the potent deep learning model including weed and seed identification, classification of plant illnesses, fruit counting, root segmentation, etc. Deep learning is a development in machine learning that successfully trains a large capacity of data, automatically picks up on the features of the input, and produces results based on predetermined rules. CNN does a good job of digesting the visual data. It is a forward computational neural systems featuring distinct input that is secret, and output levels. from one another. The convolutional layer, the pooling layer, the normalisation layer, and the fully connected layer make up the hidden layer. It also comprises a set of automatically learnable parameters (weights) through which it A system of computer algorithms known as "machine learning" is capable of learning from experience and improving itself without having explicit programmingIt is a subset of AI, use statistical tools alongside data to forecast possible outcomes. utilised to generate actionable insights.

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Making recommendations is a shared machine learning problem. All Netflix Users with accounts are given offers according on the user's prior viewing history. Unsupervised learning is being used by tech companies to enhance user experience with personalised recommendations.

YingbinZo[1] Over the past five decades, China's More than three times as much rice has been produced, mostly due to higher yields of grain than to larger cultivation areas. This expansion was previously linked to the invention of high-yielding varieties and enhanced crop management techniques including irrigation and nitrogen fertilisation. However, China has seen a plateau in rice yield during the last ten yearsJapan has to make 20% more grain by 2030 to meet its own hunger as the economy expands, assuming that annual use of rice stays constant. Raising overall rice output in a sustainable manner is a challenging endeavor due to several trends and problems in the Chinese rice farming system. Any of the primary concerns is a decrease in farmland.

Jiang, Feng, and others.[2] Identification and forecasting of diseases affecting rice leaves has always been a research priority in the area of agricultural informatics. Currently, the fields of pattern recognition are popular fields of study, as

supports vector machinery (SVM) technology. When combined, they might not only resolve the problem efficiently but also improve the accuracy of the assessment. Using convolution neural networks (CNNs), we first extract the distinctive characteristics of rice leaf illness from the pictures in this study. Next, the specific ailment is categorized and predicted using the SVM technique. The optimal parameters for every SVM model are ascertained using the ten-fold cross-validation method. As stated by the experimental findings, when the penalty parameter is The average correct recognition rate of the rice illness recognition Using an internal value of $g = 50$, the CNN and SVM model has a 96.8% accuracy rate. The precision achieved is higher than traditional replication neural net models. This work presents a novel strategy for the ongoing examination of agricultural diseases by using deep learning.

Onyejgb, L. N.[3] The methods used to identify, measure, and categorise plant diseases from digital photographs in the visible spectrum are surveyed utilizing an image online in this piece processing techniques. Although disease symptoms might appear anywhere on a plant, only approaches that meet on the outwardly evident symptoms in leaves and stems were taken into consideration. to cut down on the paper's length and because techniques for working with roots, seeds, and fruits have certain exceptional appearances that call for a more in-depth analysis. According to their objectives, the chosen ideas are categorised into three groups: detection, severity quantification, and categorization. According to the primary technological solution applied in the algorithm, each of those classes is further separated. This essay is anticipated to.

Sil Jaya[4] Machine vision techniques are widely used in agricultural research, and they have significant potential, particularly in the field of plant protection, which ultimately leads to crop management. The study outlines a software prototype approach for identifying illness in rice by the use of images of different diseased rice plants. Infected grain crops are photographed using digital photography equipment, and the images are edited. utilising image-growing and image-segmentation techniques to detect the sick plant portions. Following that, a neural network was utilised to classify the affected piece of the leaf. The techniques developed in this system are used to a variety of sick rice plants and include both image processing and soft computing.

In this study, Zhang[5] Wei Support vector machine (SVM) is discussed as a means for identifying illnesses of cucumber leaves. A new experimental programme It was proposed that, as opposed to using every leaf, utilizes every region of leaf like an experiment because of the smallest number of samples. Radial Basis Function (RBF), polynomial, and sigmoid kernel functions were also utilised in the trials to do comparison assessments. The findings demonstrated that, for classifying cucumber leaf illnesses, the SVM method based on RBF kernel function and using each spot as a sample produced the best results.

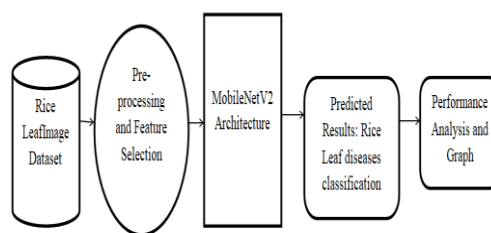


Fig. 1. Proposed Architecture

III. EXISTING MODEL

The current technique for predicting rice leaf disease makes use of a classifier called a quadratic support vector machine (SVM). A well-liked The SVM way of machine learning can be used to classification jobs. In this method, a dataset of photos of rice leaves with various diseases on them is utilized for SVM classifier training. The classification algorithm uses a set of characteristics extracted from the photos.



⌊ The current system's Quadratic SVM output had an AUC of 0.92 and an accuracy of 81.8%. As noise reduction techniques were not applied to the dataset used during the preprocessing stage, the accuracy is a little low. The set also contains an unbalanced number of photos showing rice leaf disease. In addition, we have only taken global.

The current system has a number of drawbacks. First, it relies on labor-intensive and error-prone manual feature extraction from the photos. Second, The accuracy of the classifier will be stuck if the features are not indicative of the underlying data. Last but not least, the current system might not be scalable because the SVM classifier might not be able to effectively handle enormous datasets.

⌊ Despite these drawbacks, it has been demonstrated that the current approach is successful in identifying the type of illness in rice leaves. There is always opportunity for improvement, though, and the proposed system seeks to solve some of the shortcomings of the current system by utilising

IV. PROPOSED METHODOLOGY

□ The dataset for the suggested system includes four different disease types: brown spot, sheath blight, rice leaf blast, and bacterial leaf blight. Bacterial Leaf Blight: When a leaf is harmed, bacteria cause the injured area to grow inches long. The colour will initially be yellow before changing to brown and dark brown. Brown spot is a disorder that affects plant leaves and is recognised by juvenile, round, brown patches on the leaves. Wheat and rice plants are primarily affected by this disease. Mature plants have reddish borders that are visible. Rice Blast: It causes leaves to develop an oval-shaped area with white dots and black edges. Leaf nodes may also exhibit symptoms. Sheath Blight - Affects the plant's stem and leaves. ⌊ The proposed approach for machine learning-based rice leaf disease prediction intends to solve some of the shortcomings of the current system. The accuracy, scalability, and adaptability of the proposed system are enhanced using deep learning techniques and a more effective design. The system makes advantage of MobileNetV2 architecture, a mobile device-optimized convolutional neural network (CNN) that is lightweight. The Python programming language and a number of libraries, including Keras, TensorFlow, etc., are applied to create the suggested system.

V. IMPLEMENTATION

Dataset: Prediction of Rice Leaf Disease's initial module We designed the technique to obtain the input dataset using machine learning. The process of collecting The first important step in actually creating a machine learning model is gathering data. This is an important stage since the amount and quality of data we are able to gather will determine how effectively the model works. The statistics be able to be gathered using a variety of techniques, such as web scraping and physical inputs. The data we need is stored in the model folder of the work in progress. Every researcher uses a set of files from the renowned Kaggle standards datasets source. A total of 1,396 images feature rice. leaves in the dataset. The dataset is cited by

Importing the necessary libraries Python will be the language we use for this. In order to build the primary model, partition the drill and scan data using Sklearn, turn photos into arrays of numbers using PIL, and use other libraries like pandas, numpy, matplotlib, and tensorflow, we must first import the appropriate libraries.

Retrieving the images: The photos will be retrieved from the dataset and converted into a format that can be utilised for both assessing and preparing the prototype in this module. This calls for reading, resizing, and normalising the pixel standards of the photos. The photos and their labels will be retrieved. The photos should then be resized to (224,224) since they all need to be the same size for recognition. Then, create a numpy array from the photos.

Dataset division: The dataset will be split into exercising and hard sets for this module. Create Train and Test subsets of the dataset. 80 percent train data, 20 percent test data. This will be done to validate the model's performance, test the model on omitted data to assess its correctness, and line the classical on a subset of the data. Create file and acid test datasets. 80 percent train data, 20 percent test data.

VI. CONCLUSIONS

In contemporary agricultural and industrial food production, computer vision and AI frameworks are increasingly integrated across various stages. Given the potential for substantial losses caused by rice hide sicknesses in the agricultural sector, these frameworks offer a promising approach to more accurately identify multiple diseases affecting rice crops. Their efficient deployment can automate tedious tasks, generating abundant data for subsequent analyses.



The endings of the project underscore the potential of deep-seated be taught processes and optimized architectures in predicting rice leaf diseases. Farmers stand to benefit significantly from this proposed system, which boasts improvements in accuracy, scalability, flexibility, user-friendliness, and efficiency compared to current methods. Moreover, the system appeals to academics studying agriculture.

By employing advanced Through improved structure and deep neural networks, the project "Rice Blade Predicting Disease Employing Ai Learning" introduces a groundbreaking approach to identifying and forecasting rice leaf diseases. The proposed system surpasses the working of existing methods, including a quadratic SVM classifier, achieving a training accuracy of 98.34% and validation accuracy of 95.21%.

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