

# Plant Disease Prediction for Sustainable Agriculture

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## Learning Objectives:

Plant Disease Knowledge Identify common plant diseases and their symptoms. Image Processing Understand image preprocessing techniques for machine learning. Machine Learning Application Apply CNNs for image classification using a pre-trained model. Streamlit Framework Build interactive web applications for user input and output. Practical Skills Integrate machine learning models with web applications. Data Management Handle file uploads and manage image storage effectively



## Tools and Technology used:

**Programming Language Python:** The primary language used for developing the application.

**Web Framework Streamlit:** A framework for building interactive web applications easily.

**Machine Learning Library TensorFlow:** A library for building and training machine learning models, specifically for the CNN used in this application.

**Image Processing Library OpenCV:** A library for image processing tasks, including reading and resizing images.

**Numerical Computing Library NumPy:** A library for numerical operations, used for handling image data.

**Image Handling PIL (Pillow):** A library for opening, manipulating, and saving image files.

**File Management OS Module:** Used for handling file paths and saving uploaded images.

## Methodology:

**Image Upload** - Users upload an image of a plant through the Streamlit interface using the file uploader component. **Image Saving** The uploaded image is saved to a specified directory on the server for further processing. **Image Preprocessing** The saved image is read using OpenCV. The image is resized to 224x224 pixels to match the input size required by the CNN model. The color format is converted from BGR to RGB. The pixel values are normalized by scaling them to a range of 0 to 1. **Model Loading** The pre-trained CNN model is loaded using TensorFlow. **Prediction** The preprocessed image is reshaped to fit the model's input requirements. The model predicts the class of the plant disease by outputting probabilities for each class. The class with the highest probability is identified using np.argmax. **Result Display** The predicted class name is retrieved from a predefined list of plant disease classes. The prediction result is displayed to the user through the Streamlit interface. **User Interaction** Users can view the uploaded image and the prediction result, enhancing the interactive experience of the application

## Problem Statement:

The agricultural sector faces significant challenges due to the prevalence of plant diseases, which can lead to reduced crop yields and economic losses. Farmers and agricultural stakeholders often lack timely and accurate methods to identify and diagnose these diseases, resulting in delayed interventions and ineffective management strategies. To address this issue, there is a need for an efficient and user-friendly system that can: Accurately predict plant diseases based on images of affected plants. Provide immediate feedback to users, enabling them to take prompt action. Utilize machine learning techniques to enhance the accuracy of disease identification. This project aims to develop a Plant Disease Prediction System that leverages convolutional neural networks (CNNs) and a web-based interface using Streamlit. The system will allow users to upload images of plants, process these images, and provide predictions regarding potential diseases, thereby supporting sustainable agricultural practices and improving crop management.

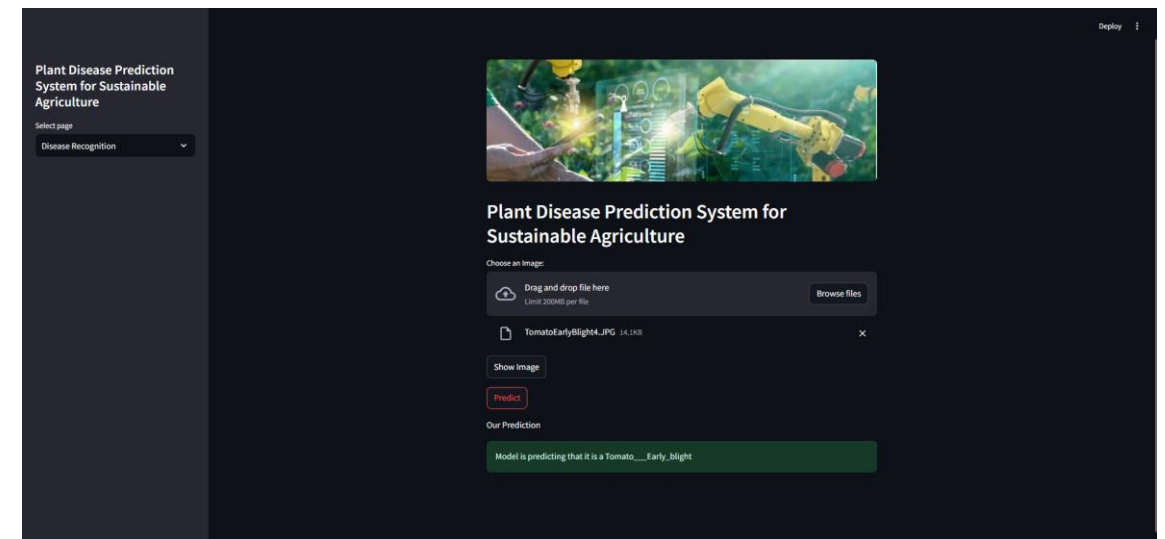
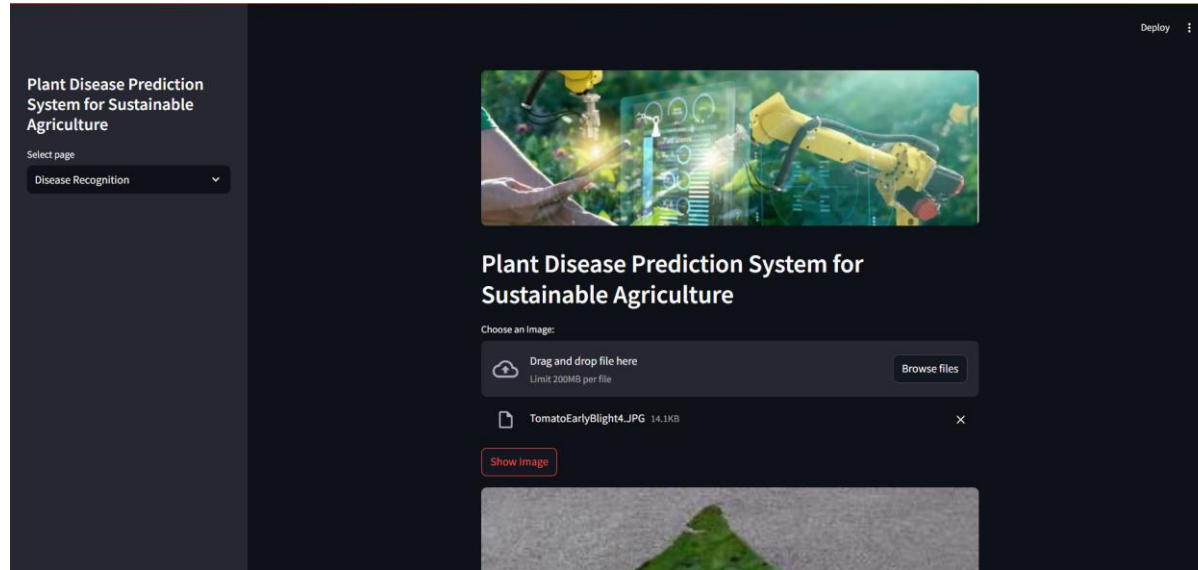
## Solution:

The proposed solution is a Plant Disease Prediction System that utilizes machine learning and web technology to provide accurate and timely predictions of plant diseases based on uploaded images. The key components of the solution are as follows:

- Image Upload Interface** A user-friendly web interface built with Streamlit allows users to easily upload images of plants for analysis.
- Image Processing Pipeline** The uploaded images are processed using OpenCV, which includes resizing, color conversion, and normalization to prepare the images for input into the machine learning model.
- Machine Learning Model** A pre-trained convolutional neural network (CNN) model is employed to analyze the processed images. The model is capable of identifying various plant diseases based on learned features from a diverse dataset.
- Prediction Mechanism** The system predicts the class of the plant disease by evaluating the processed image through the CNN model. The prediction is based on the highest probability output from the model.
- Result Presentation** The predicted disease class is displayed to the user along with relevant information, enhancing the user's understanding of the diagnosis and potential management strategies.
- Interactive Features** Users can view the uploaded image alongside the prediction results, fostering an engaging and informative experience.

**Sustainable Agriculture Support** By providing quick and accurate disease identification, the system empowers farmers and agricultural stakeholders to make informed decisions, ultimately contributing to sustainable agricultural practices and improved crop health. This solution aims to bridge the gap between technology and agriculture, facilitating better disease management and promoting sustainable farming practices.

## Screenshot of Output:





## Conclusion:

The Plant Disease Prediction System developed in this project effectively addresses the critical challenge of timely and accurate identification of plant diseases in agriculture. By leveraging advanced machine learning techniques, specifically convolutional neural networks, and integrating them into a user-friendly web application using Streamlit, the system provides a practical solution for farmers and agricultural stakeholders. Key takeaways from the project include:

- Enhanced Decision-Making:** The system enables users to quickly diagnose plant diseases, allowing for prompt interventions that can mitigate crop losses and improve overall agricultural productivity.
- User Engagement:** The interactive interface fosters user engagement, making it easier for individuals with varying levels of technical expertise to utilize the technology effectively.
- Support for Sustainable Practices:** By facilitating accurate disease identification, the system contributes to sustainable agricultural practices, promoting healthier crops and reducing the reliance on chemical treatments.
- Future Potential:** The framework established in this project can be expanded to include additional features, such as disease management recommendations, integration with agricultural databases, and support for a wider range of plant species.

In summary, the Plant Disease Prediction System not only demonstrates the potential of machine learning in agriculture but also serves as a valuable tool for enhancing crop management and supporting sustainable farming practices. The successful implementation of this system paves the way for further innovations in agricultural technology, ultimately benefiting farmers and the agricultural industry as a whole.