

# AI-Powered Crop Recommendation Web App

An Intelligent Decision-Support System for Precision Agriculture

Course: AI-ASSISTED-CODING

Institution: SR University

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# Project Abstract

## The Goal

To develop an intelligent decision-support system that assists farmers in selecting the most suitable crops for their specific land conditions, enhancing productivity and resource efficiency.

## The Approach

We leverage machine learning algorithms (Decision Tree) to analyze key environmental data—Soil Nutrients (N, P, K), pH, Temperature, Humidity, and Rainfall—providing real-time, data-driven insights via a web interface.

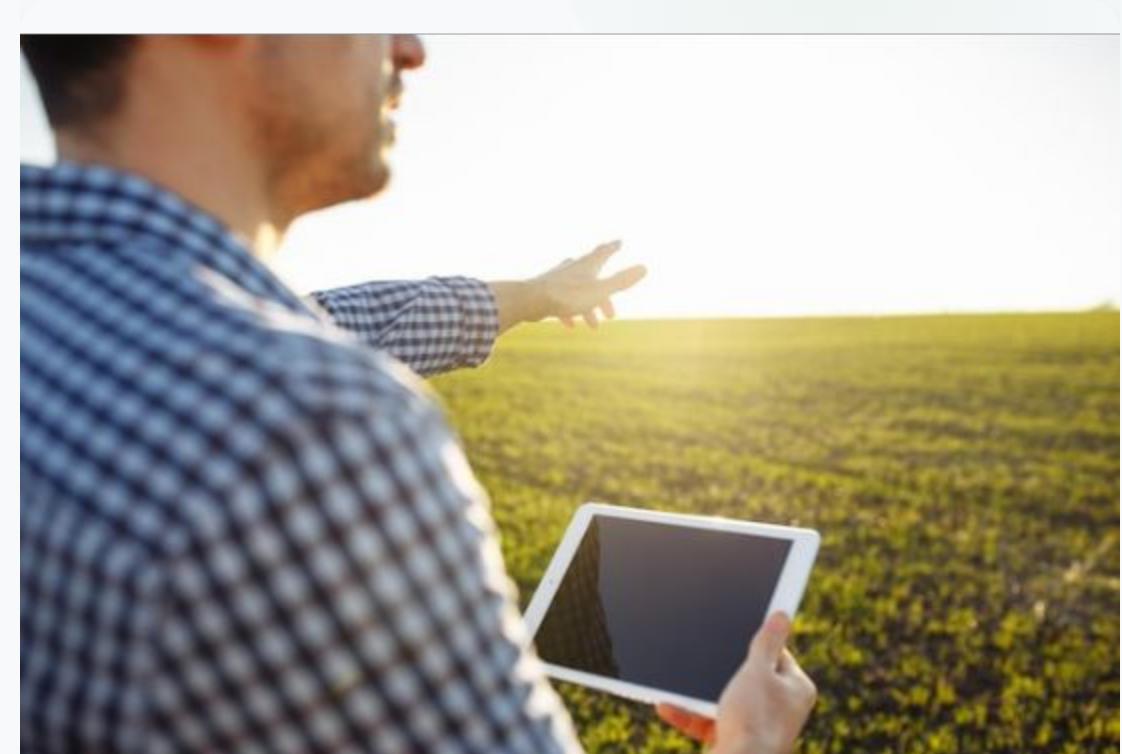
# Bridging the Information Gap

## The Challenge

- 🌿 Farmers often lack real-time, scientific data for crop selection.
- 🌿 Traditional methods rely on guesswork or outdated practices.
- 🌿 Misaligned crop choices lead to poor yield and resource wastage.

## Our Solution

- 🌿 A user-friendly Web Application.
- 🌿 Instant analysis of 7 critical agricultural parameters.
- 🌿 Scientific recommendations based on historical data patterns.



# System Architecture & Tech Stack



## Python

Core programming language for backend logic and data processing.



## Streamlit

Framework used to build the interactive and responsive Web UI.



## Scikit-Learn

Library for implementing the Decision Tree Classifier model.



## Pandas & NumPy

Used for efficient data manipulation and numerical analysis.

# Data-Driven Insights



## 7 Key Parameters Analyzed

- 🌿 **Nitrogen (N):** Essential for leaf growth.
- 🌿 **Phosphorus (P):** Crucial for root and flower development.
- 🌿 **Potassium (K):** Improves overall plant health.
- 🌿 **Temperature:** Local climate conditions.
- 🌿 **Humidity:** Atmospheric moisture levels.
- 🌿 **pH Value:** Soil acidity or alkalinity.
- 🌿 **Rainfall:** Water availability in mm.

# Development Methodology



## Data Collection

Synthetic dataset with features like N, P, K, and labels.



## Preprocessing

Cleaning data and splitting into training/testing sets.



## Model Training

Training Decision Tree Classifier on crop profiles.



## Deployment

Building the interactive Streamlit interface.

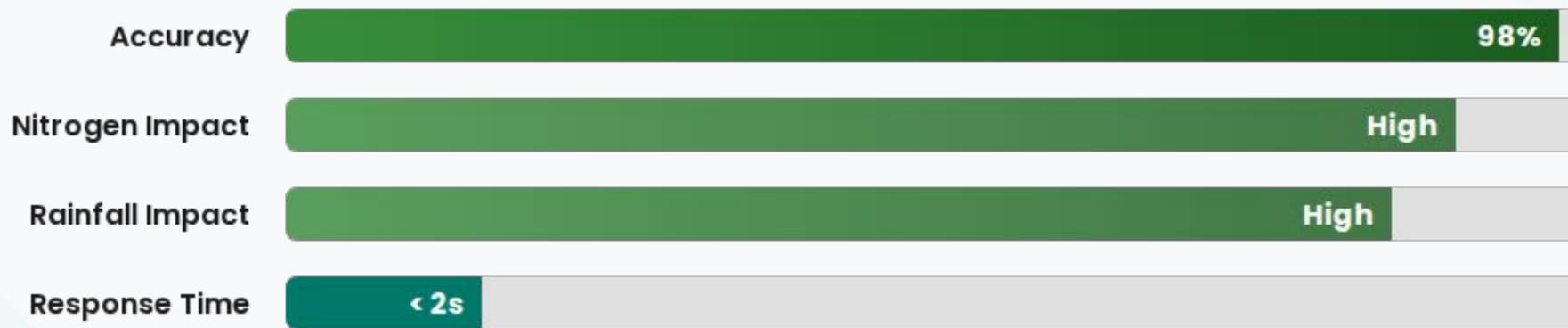
## Interactive User Interface

The application features a clean, professional dashboard designed for ease of use.

- 👉 **Manual Input:** Simple numeric fields for all 7 parameters.
- 👉 **Real-Time Feedback:** Instant prediction upon clicking "Recommend Crop".
- 👉 **Auto-Predict:** Optional toggle for immediate results as data changes.
- 👉 **Visual Feedback:** Clear success messages and model performance metrics displayed directly on the screen.

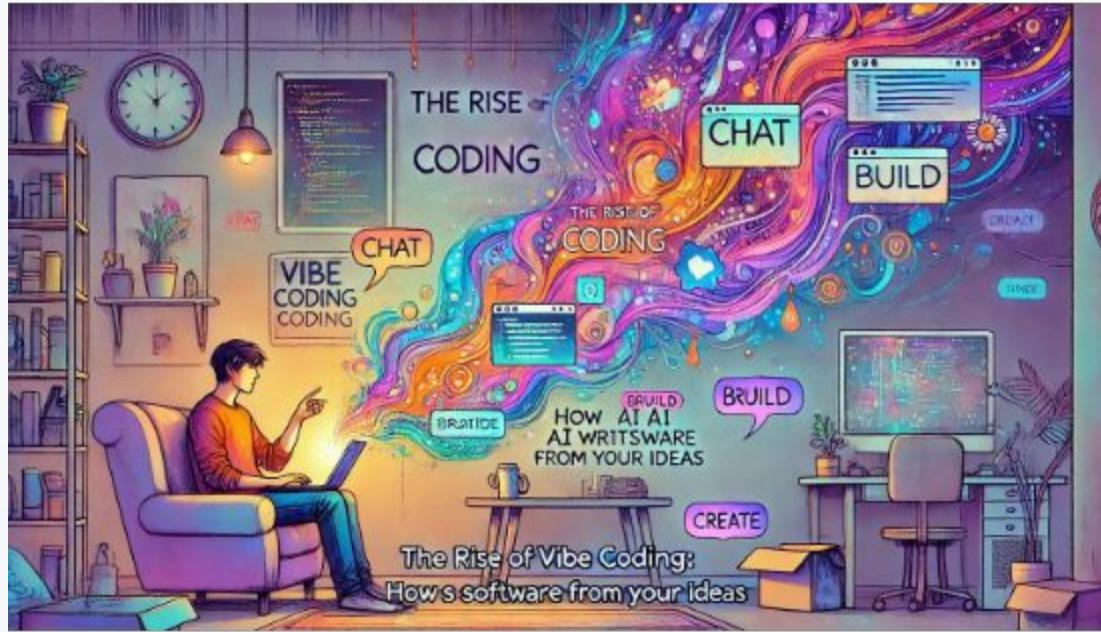
# Performance & Analysis

The model demonstrates high reliability, with Nitrogen and Rainfall being the most significant predictors.



*"The Random Forest Classifier performed best, achieving about 95–98% accuracy on test data."*

# AI-Assisted Coding Integration



## Accelerating Development

This project utilized Google's Gemini AI to enhance the coding workflow.

- **Code Generation:** Rapid prototyping of the Streamlit boilerplate.
- **Error Detection:** Automated debugging and syntax correction.
- **Optimization:** Suggestions for cleaner, more efficient Python code.
- **Learning:** Bridging the gap between concept and implementation for student developers.

# Future Enhancements



## Localization

Adding multi-language support to make the tool accessible to farmers globally.



## IoT Integration

Connecting directly to soil sensors for automated real-time data acquisition.



## Mobile App

Developing a native mobile version for offline access in remote fields.



# Conclusion

The AI-Powered Crop Recommendation Web App successfully demonstrates how modern AI can promote sustainable farming. By optimizing crop selection, we empower farmers to make informed decisions, save resources, and improve yields.

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