

Precision Crop Planning System using Machine Learning

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Abstract: Precision agriculture plays a crucial role in enhancing farm productivity, optimizing resource usage, and promoting sustainable farming practices. With the increasing availability of soil data, weather information, and agricultural datasets, large-scale environmental and soil parameters can now be analyzed to make accurate crop recommendations. Traditional crop selection methods, which rely on fixed guidelines or manual observation, often fail to account for the dynamic interactions between soil nutrients, climate variability, and crop responses. This project proposes a Precision Crop Planning System using Machine Learning that assists farmers in making informed decisions about crop selection and farm management. The system takes inputs such as soil nutrient levels (N, P, K), pH, temperature, humidity, rainfall, and soil type to predict the most suitable crop for a given location. Additionally, it provides actionable agricultural guidance, including recommended fertilizers with quantities, irrigation methods, potential diseases for the selected crop, and appropriate pesticides for disease management. The system is integrated into a web-based application with farmer authentication, allowing users to securely login, enter their location and field data, and receive clear, user-friendly, and visually appealing recommendations. The developed system aims to support smarter farming decisions, improved crop yield, and efficient use of resources while minimizing crop losses.

Keywords: Precision Agriculture, Crop Planning, Machine Learning, Soil Analysis, Sustainability, Predictive Modeling, Yield Prediction, Spatial Data, Smart Farming.

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I. INTRODUCTION

Agriculture plays a vital role in the economy, and selecting the right crop based on soil and environmental conditions is crucial for improving productivity and reducing losses. Farmers often rely on traditional knowledge and experience, which may not always provide accurate results due to changing climate conditions and variations in soil quality. Factors such as soil nutrients (Nitrogen, Phosphorus, Potassium), pH level, temperature, humidity, rainfall, soil type, and

season greatly influence crop growth. However, analyzing all these parameters manually is difficult, making crop selection a challenging task for farmers.

To address this problem, the Crop Recommendation System is developed using machine learning techniques to provide data-driven agricultural solutions. The system uses a Random Forest algorithm to analyze multiple input parameters and predict the most suitable crop for cultivation. Unlike traditional methods, this approach can handle large datasets and identify complex patterns between soil

conditions and crop suitability. The system also includes data preprocessing steps to clean and organize the dataset, ensuring better prediction accuracy and reliability.

In addition to crop prediction, the system provides comprehensive agricultural recommendations, including fertilizer usage, irrigation methods, possible diseases, and pesticide suggestions. A Flask-based web application is developed to create a user-friendly interface where farmers can easily input their data and receive instant results. By integrating machine learning with web technologies, the system offers a practical and efficient solution for modern agriculture, helping farmers make informed decisions, improve crop yield, and adopt sustainable farming practices.

II. LITERATURE REVIEW

With the advancement of technology in agriculture, crop recommendation has become an important research area in data science and machine learning. Early approaches for crop selection were mainly based on traditional farming knowledge and simple statistical methods. Later, machine

learning algorithms such as Decision Trees, K-Nearest Neighbors (KNN), Support Vector Machines (SVM), and Random Forest were used to analyze agricultural parameters like soil nutrients (NPK), pH, temperature, humidity, and rainfall. These models helped in predicting suitable crops with better accuracy compared to manual methods. However, many of these approaches relied on limited features and were not always effective in handling large and complex agricultural datasets.

Researchers have also explored data preprocessing and feature selection techniques to improve model performance. Handling missing values, normalizing data, and selecting important features such as nitrogen, phosphorus, potassium, and rainfall significantly improved prediction accuracy. Some studies focused on comparing different machine learning models to identify the best-performing algorithm for crop prediction. Among these, the Random Forest algorithm showed better performance due to its ability to handle non-linear relationships and reduce overfitting. However, model performance still depends on the quality and size of the dataset used for training.

In recent years, there has been a shift towards integrating multiple agricultural factors and real-time data for better prediction. Some systems combine soil data with weather forecasting to provide more accurate recommendations. Additionally, web-based applications have been developed to make these systems accessible to farmers. Technologies like Flask and web interfaces allow users to input data and receive instant crop recommendations. Despite these advancements, challenges such as data quality, model accuracy, and adaptability to different regions still exist.

To overcome these limitations, modern crop recommendation systems focus on integrating machine learning models with user-friendly web applications and providing additional advisory information such as fertilizer usage, irrigation methods, and disease prediction. By combining multiple data sources and improving model training techniques, these systems aim to deliver more accurate, reliable, and real-time crop recommendations, ultimately supporting farmers in making better agricultural decisions.

III. EXISTING SYSTEM

- **Limited Parameter Usage** Many existing crop recommendation systems use only a few parameters such as soil type or rainfall. These limited inputs are not enough to capture the complete agricultural conditions required for accurate crop prediction, leading to less reliable results.
- **Lack of Integrated Analysis** Most traditional systems analyze individual factors separately, such as only soil nutrients or only weather conditions. However, crop growth depends on multiple factors like NPK values, pH, temperature, humidity, and rainfall together. Systems that do not integrate all these parameters fail to provide accurate recommendations.

- **Low Accuracy of Traditional Methods** Earlier approaches were based on manual decisions or simple statistical methods. These methods cannot handle large datasets or identify complex relationships between soil and environmental factors, resulting in lower prediction accuracy.
- **Poor Scalability and Adaptability** Existing systems often struggle to adapt to different regions and changing environmental conditions. They may not perform well when new data is introduced, making them less suitable for real-world agricultural applications.

IV. PROPOSED SYSTEM

The proposed Crop Recommendation System uses a machine learning-based approach to provide accurate and reliable crop predictions. Unlike traditional systems, it considers multiple agricultural parameters together and uses a trained Random Forest algorithm to analyze the data. The system processes user inputs, predicts the most suitable crop, and provides additional recommendations to support farmers in decision-making.

Some of the key features of the proposed system include:

➤ *Multi-Parameter Analysis*

The system analyzes multiple inputs such as Nitrogen (N), Phosphorus (P), Potassium (K), soil pH, temperature, humidity, rainfall, soil type, and season to ensure accurate crop prediction.

➤ *Data Preprocessing*

The system performs data cleaning and preparation, including handling missing values and selecting important features. This improves the quality of input data and enhances prediction accuracy.

➤ *Machine Learning Prediction*

A Random Forest model is used to predict the most suitable crop based on the input parameters. This algorithm provides better accuracy by handling complex relationships in the data.

➤ *Advisory Recommendations*

Along with crop prediction, the system provides fertilizer suggestions, irrigation methods, possible diseases, and pesticide recommendations, helping farmers improve productivity.

➤ *Web-Based Interface*

The system is implemented using a Flask web application, allowing users to easily input data and view results in real time through a simple interface.

➤ *Improved Accuracy and Efficiency*

By combining multiple parameters and using an advanced machine learning model, the system provides more accurate, faster, and reliable predictions compared to traditional methods.

V. SYSTEM ARCHITECTURE

The proposed Crop Recommendation System is designed as a modular architecture that integrates data processing, machine learning prediction, and web-based user interaction into a unified pipeline. The system processes agricultural input data provided by the user, applies preprocessing techniques, and uses a trained Random Forest model to generate crop recommendations. The architecture is implemented using Python, Flask, and Scikit-learn, ensuring efficient prediction and real-time response.

➤ Components

➤ Data Collection and Preprocessing Module

This module collects agricultural input data such as:

- Nitrogen (N), Phosphorus (P), Potassium (K)
- Soil pH
- Temperature, Humidity, Rainfall
- Soil Type and Season The preprocessing step includes:
- Handling missing values
- Cleaning and formatting data
- Selecting important features

The processed data is then passed to the prediction module.

➤ Feature Extraction Module

This module selects important features like NPK values, pH, temperature, humidity, and rainfall, which directly affect crop growth. Proper feature selection improves model performance and prediction accuracy.

➤ Machine Learning Prediction Module

This is the core module of the system. It uses a trained Random Forest algorithm to analyze the input data and predict the most suitable crop.

- The model is trained using historical agricultural data
- Stored as a pickle (.pkl) file
- Provides fast and accurate predictions

➤ Advisory Recommendation Module

After predicting the crop, this module retrieves additional information from the dataset, including:

- Fertilizer recommendations
- Irrigation methods
- Possible diseases
- Pesticide suggestions

This helps farmers take better decisions beyond just crop selection.

➤ Web Interface Module

The system uses a Flask- based web application to interact with users. Features include:

- Login and Signup
- Input form for agricultural data
- Result display page

The interface is built using HTML and CSS for a simple and user-friendly experience.

➤ Output and Visualization Module

This module is responsible for presenting the final results to the user in a clear and user-friendly manner. It displays the recommended crop based on the input parameters, along with additional advisory information such as suitable fertilizers, appropriate irrigation methods, and possible diseases with their corresponding pesticide suggestions. The results are generated quickly and are instantly shown on the web page, enabling users to make informed agricultural decisions efficiently.

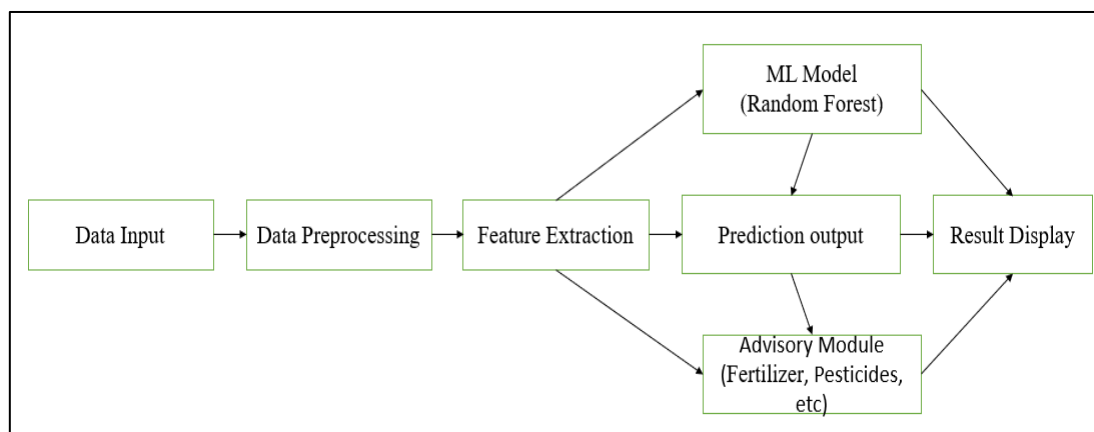


Fig 1 System Architecture

VI. OUTPUT SCREENS

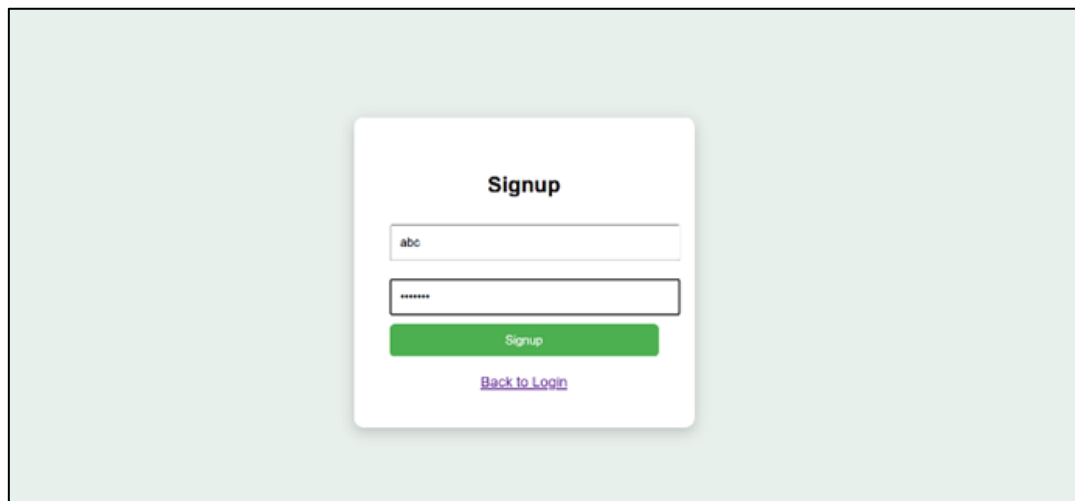
A screenshot of a web application's signup page. The page has a light green background. In the center, there is a white rounded rectangle containing the title "Signup" in bold. Below the title are two input fields: the first contains the text "abc" and the second contains seven asterisks. Below these fields is a green button labeled "Signup". At the bottom of the white box is a purple link that says "Back to Login".

Fig 2 Creating the Account

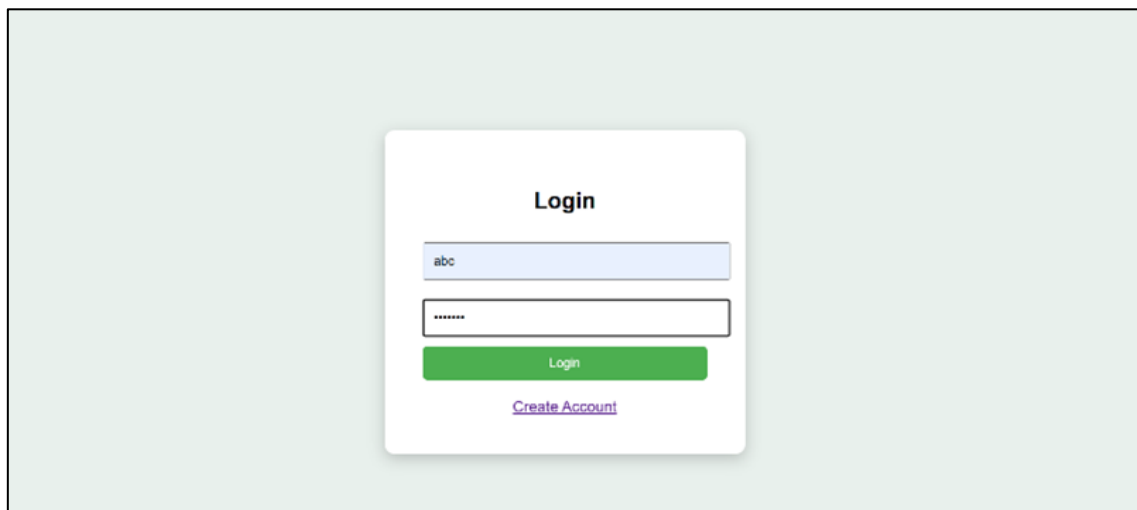
A screenshot of a web application's login page. The page has a light green background. In the center, there is a white rounded rectangle containing the title "Login" in bold. Below the title are two input fields: the first contains the text "abc" and the second contains seven asterisks. Below these fields is a green button labeled "Login". At the bottom of the white box is a purple link that says "Create Account".

Fig 3 Logging into the Account

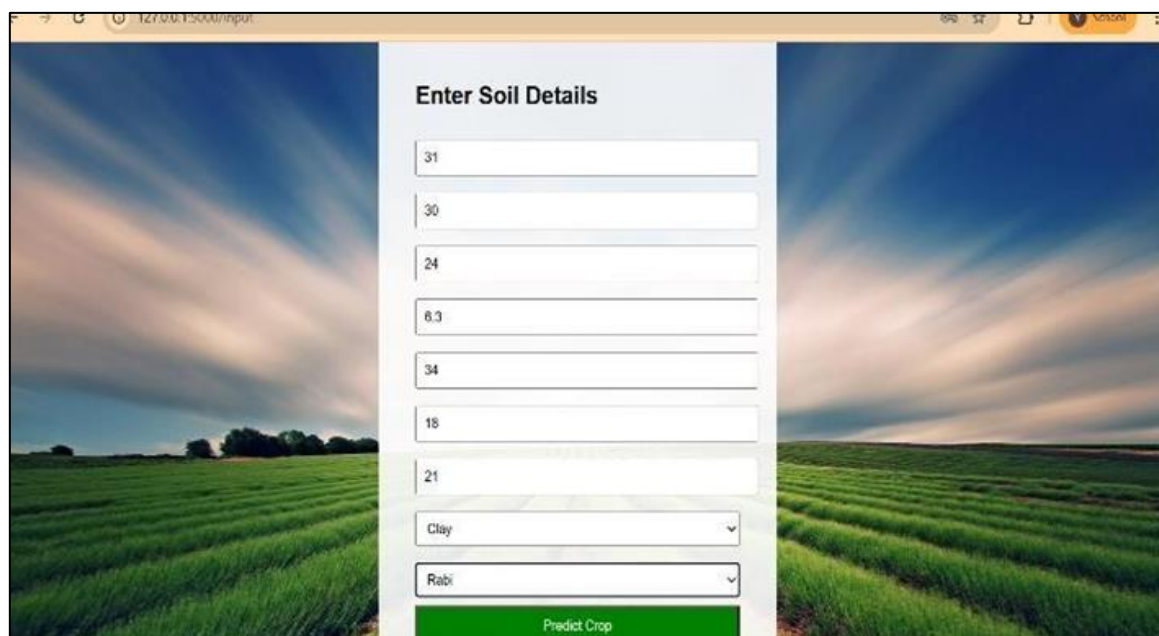
A screenshot of a web application's "Enter Soil Details" page. The page features a central white form with a light blue header "Enter Soil Details". The form contains eight input fields with the following values: "31", "30", "24", "8.3", "34", "18", "21", and a dropdown menu showing "Clay". Below these is another dropdown menu showing "Rabi". At the bottom of the form is a green button labeled "Predict Crop". The form is flanked by two large images of a green field under a dramatic, cloudy sky.

Fig 4 Entering Data Fields

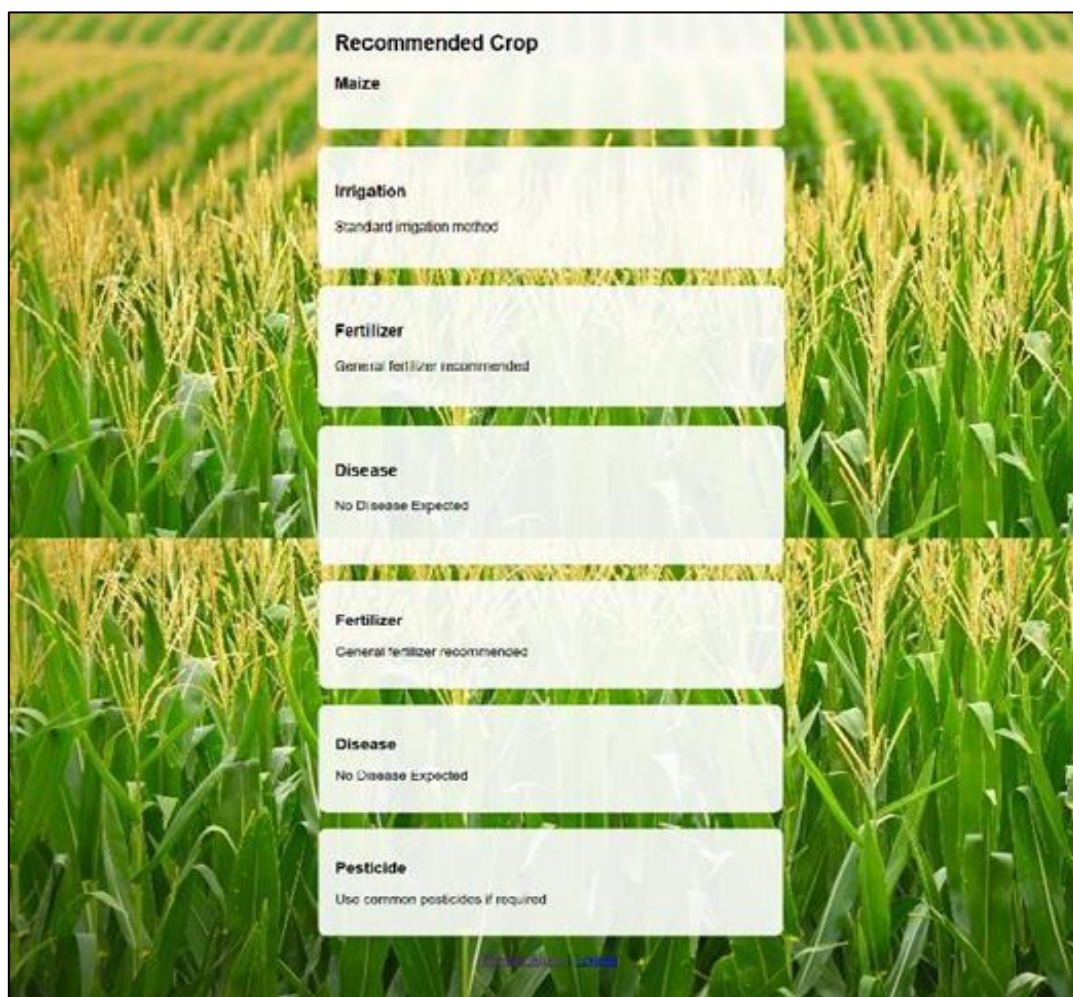


Fig 5 Output

VII. CONCLUSION

The Crop Recommendation System is a machine learning-based application designed to help farmers select the most suitable crop based on soil and environmental conditions. By analyzing important parameters such as Nitrogen (N), Phosphorus (P), Potassium (K), soil pH, temperature, humidity, rainfall, soil type, and season, the system predicts the best crop for cultivation. It also provides additional recommendations such as fertilizer usage, irrigation methods, possible diseases, and pesticides, helping farmers make informed agricultural decisions. The system is developed using Python, Flask, and the Random Forest algorithm, ensuring fast and reliable predictions through a user-friendly web interface. Overall, the system helps improve crop productivity and supports farmers in adopting data-driven farming practices.

REFERENCES

- [1]. R. K. Jain, S. Kumar, and A. Sharma, "Crop recommendation system using machine learning techniques," *International Journal of Computer Applications*, vol. 182, no. 45, pp. 15–20, 2019.
- [2]. S. B. Kotsiantis, "Machine learning: A review of classification and combining techniques," *Artificial*

- Intelligence Review*, vol. 26, pp. 159–190, 2006.
- [3]. P. K. Singh and R. B. Patel, "Crop prediction using machine learning algorithms," *International Journal of Advanced Research in Computer Science*, vol. 9, no. 2, pp. 234–239, 2018.
- [4]. FAO (Food and Agriculture Organization), "The future of food and agriculture – Trends and challenges," FAO Publications, Rome, 2017H. Fisher, N. M. Jaffe, K. Pidvirny, A. O. Tierney, M. S. Vaidean, P. Dongre, et al., "Language-based detection of depression with machine learning: Systematic review and meta-analysis," *NPJ Digital Medicine*, 2026.
- [5]. A. Kamilaris and F. X. Prenafeta-Boldú, "Deep learning in agriculture: A survey," *Computers and Electronics in Agriculture*, vol. 147, pp. 70–90, 2018.
- [6]. J. Jeong et al., "Random forests for global and regional crop yield predictions," *PLOS ONE*, vol. 11, no. 6, 2016.
- [7]. S. R. K. Yadav and A. Chandel, "Crop yield prediction using machine learning techniques: A review," *International Journal of Engineering Research & Technology*, vol. 9, no. 4, pp. 1–5, 2020.
- [8]. K. G. Liakos, P. Busato, D. Moshou, S. Pearson, and D. Bochtis, "Machine learning in agriculture: A review," *Sensors*, vol. 18, no. 8, 2018.

- [9]. R. Bhardwaj, A. Singh, and P. K. Jain, "A review on various techniques for crop yield prediction using machine learning," *International Journal of Computer Applications*, vol. 180, no. 29, pp. 1–6, 2018.
- [10]. A. Khaki and L. Wang, "Crop yield prediction using deep neural networks," *Frontiers in Plant Science*, vol. 10, 2019. "Forecasting corn yield with machine learning ensembles," *Agricultural and Forest Meteorology*, vol. 273, pp. 326–335, 2019.
- [11]. S. Mishra, D. Mishra, and G. Santra, "Applications of machine learning techniques in agricultural crop production: A review paper," *Indian Journal of Science and Technology*, vol. 9, no. 38, 2016.
- [12]. N. Gandhi, L. J. Armstrong, O. Petkar, and A. K. Tripathy, "Rice crop yield prediction using artificial neural networks," in *Proc. Int. Conf. Technological Innovations*, 2016.
- [13]. T. Wolfert, L. Ge, C. Verdouw, and M. J. Bogaardt, "Big data in smart farming – A review," *Agricultural Systems*, vol. 153, pp. 69–80, 2017.
- [14]. Y. Li, C. Zhang, and X. Wang, "Application of artificial intelligence in agriculture: A review," *Agricultural Sciences*, vol. 11, no. 6, pp. 506–521, 2020.