

# Precision Agriculture: Leveraging Sensor Technology for Soil Analysis to Recommend Crop Using Random Forest Approach

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**Abstract**—In modern agriculture, leveraging advanced technology can significantly enhance crop cultivation practices and optimize yields. This study proposes a comprehensive methodology for the development of a Crop suggestion system aimed at assisting farmers in making informed decisions regarding crop selection based on real-time sensor data analysis, machine learning techniques, and intuitive voice guidance. The objectives encompass sensor deployment for monitoring key parameters such as Nitrogen, Phosphorus, and Potassium(NPK) levels, soil moisture, humidity, and temperature, followed by data collection, preprocessing and feature engineering to extract relevant insights from the collected data. Machine learning models are then developed and trained using historical sensor data to predict crop suitability for specific environmental conditions. A crop recommendation system is designed to integrate with real-time sensor data, providing tailored suggestions for optimal crop selection based on current environmental conditions. A user-friendly interface is developed, incorporating voice guidance capabilities to facilitate seamless interaction with the system. By following this methodology, farmers can make data-driven decisions to enhance agricultural productivity and sustainability, ultimately leading to improved crop yields and economic outcomes.

**Keywords**—modern agriculture, NPK levels, soil moisture, humidity, temperature, data collection, preprocessing, feature engineering, crop recommendation system.

## 1. Introduction

In contemporary agriculture, the integration of advanced technologies has become indispensable for optimizing crop production and ensuring sustainable agricultural practices. A key area of focus lies in leveraging sensor technologies coupled with machine

learning algorithms to provide real-time insights into soil and environmental conditions. By analyzing vital parameters such as Nitrogen, Phosphorus, and Potassium(NPK) levels, soil moisture, humidity, and temperature, farmers can make informed decisions regarding crop selection and cultivation practices. This chapter presents a comprehensive methodology for the development of a Crop Suggestion System aimed at assisting farmers in optimizing crop selection and enhancing agricultural productivity.

The methodology outlined herein encompasses sensor deployment, data collection, preprocessing, feature engineering, machine learning model development, and the creation of a crop recommendation engine. Drawing from recent advancements in sensor technology and machine learning algorithms, this approach enables the extraction of actionable insights from real-time

data, facilitating timely decision-making for farmers. Moreover, the incorporation of voice guidance capabilities in both English and regional languages enhances accessibility and usability, ensuring that the system caters to a diverse user base.

By addressing the objectives of conducting soil analysis, predicting crop suitability, and delivering information through a user-friendly interface with voice assistance, this research aims to bridge the gap between traditional agricultural practices and modern technological solutions. By amalgamating insights from the latest literature and emerging opportunities in the field, this study seeks to contribute to the advancement of precision agriculture and sustainable farming practices. The subsequent chapters of this report will delve deeper into the methodology, implementation details, results, and implications of the Crop Suggestion System, thereby providing a comprehensive understanding of its potential impact on agricultural productivity and sustainability.

## 2. Objectives

1. Conduct analysis on mud samples to determine the NPK (Nitrogen, Phosphorus, and Potassium) content, soil moisture levels, humidity, and temperature.
2. Utilize the values in finding soil type to predict the most suitable crop for cultivation.
3. Present the necessary information through a display interface and provide voice assistance in both English and a trained regional language, such as Kannada.

## 3. Methods

### 3.1. Sensor Deployment:

Install sensors in the farmland to monitor key parameters such as NPK (Nitrogen, Phosphorus, Potassium) levels, soil moisture, humidity, and temperature. These sensors can be deployed at strategic locations throughout the field to capture variability in soil and environmental conditions.

#### 3.1.1. Data Collection and Preprocessing:

- Collect real-time sensor data from the field and store it in a centralized database. Ensure that the data is time-stamped and synchronized for accurate analysis.
- Preprocess the data to remove noise, outliers, and missing values. Normalize or standardize the data to ensure consistency and comparability across different sensors and parameters.

### 3.2. Feature Engineering:

- Extract relevant features from the sensor data that are indicative of soil and environmental conditions conducive to crop growth. This may include aggregating data over time intervals, calculating derived features such as average temperature or cumulative rainfall, and incorporating external data sources such as weather forecasts.

### 3.3. Machine Learning Model Development:

- Train machine learning models using historical sensor data and corresponding crop yield or performance data. Consider using techniques such as regression, classification, or clustering, depending on the specific objectives of the system.
- Experiment with different algorithms such as

decision trees, random forests, support vector machines, or neural networks to find the most suitable model for predicting crop suitability based on sensor data.

### 3.4. Model Evaluation and Validation:

- Evaluate the performance of the trained models using appropriate metrics such as accuracy, precision, recall, F1-score, or mean squared error.
- Validate the models using cross-validation techniques to ensure robustness and generalizability across different regions and growing conditions.

### 3.5. Crop Recommendation Engine:

- Develop a recommendation engine that integrates the trained machine learning models with the real-time sensor data.
- Based on the current sensor readings and historical data, the recommendation engine should suggest the most suitable crops for planting in a given area, taking into account factors such as soil nutrient levels, moisture content, temperature, and humidity.

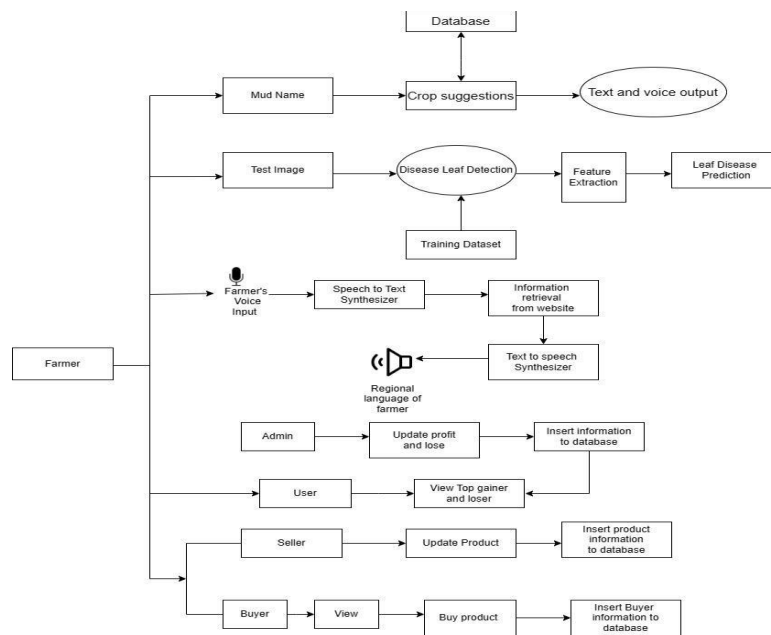
### 3.6. User Interface and Voice Guidance:

- Design a user-friendly interface for farmers to interact with the system. This could be a mobile app, web dashboard, or dedicated hardware device.
- Implement voice guidance capabilities to provide spoken instructions and recommendations to users. Integrate speech recognition and synthesis technologies to enable natural language interaction with the system.

### 3.7. Field Testing and Iterative Improvement:

- Deploy the Crop Suggestion System in real-world farming scenarios and gather feedback from users.
- Continuously monitor the system's performance and collect additional data to further train and refine the machine learning models.
- Iterate on the system based on user feedback and emerging technological advancements to enhance accuracy, usability, and reliability over time.

#### 4. Architecture

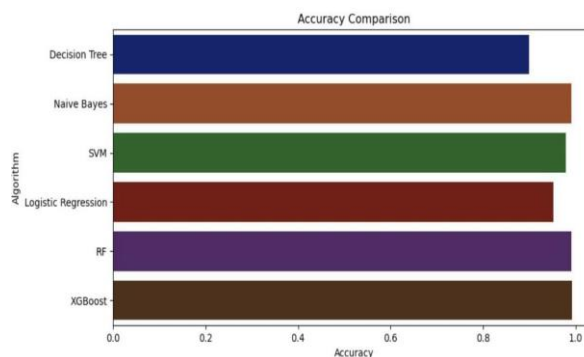


**Fig 1 . Complete System Architecture**

As seen in Fig 1. the procedure for speech recognition involves capturing and processing audio signals, extracting relevant features, and utilizing acoustic and language models to translate spoken language to written language. In parallel, the GTTS (Google Text-to-Speech) module transforms text into synthesized speech through pre-processing, text normalization, linguistic analysis, prosodic prediction, and waveform generation. Convolutional neural networks (CNNs) play a vital function in image data processing, using convolution and pooling layers to realize patterns effectively and classify images. Transitioning to an agricultural context, an e-commerce application

streamlines the farmer-consumer interaction, enabling farmers to submit product information, which are then accessible to users for viewing and purchasing. The system incorporates features of an e-commerce website, providing a user-friendly interface for farmers to upload, manage, and sell their products, contributing to a seamless and efficient marketplace experience. Additionally, the system analyzes profit and loss data, displaying information on top gainers and losers, offering valuable insights to both buyers and sellers. A data flow diagram visualizes the information flow in the system, providing an outline of data processing.

#### 5. Results



**Fig 2. Comparison of accuracy**

In Fig 2. The algorithms listed on the y-axis are decision tree, Naive Bayes, SVM, logistic regression, random forest, and XGBoost. The x-axis shows the accuracy, ranging from 0.0 to 1.0. It appears that

the decision tree model has the lowest accuracy, followed by Logistic Regression and SVM. RF and XGBoost appear to have the highest accuracy among the six algorithms listed.

| Model                                       | %Accuracy |
|---|-----------|
| Shakib Mahmud Dipto[4]<br>(SVM)             | 98        |
| Sita Rani [6]<br>(Random forest classifier) | 96.4      |
| Dhruvi Gosai [7]<br>(RF)                    | 97.8      |
| Maaz Patel [8]<br>(Neural network)          | 91        |
| Proposed System(RF)                         | 99.09     |

**Table 1: Comparative Analysis of different models**

Table 1 displays the accuracy percentage for various methods. The suggested architecture provides a higher accuracy rate than the other models and architectures listed above, as the table illustrates. Thus, reducing error rates can be accomplished with the help of the suggested model.

**6. Discussions**

Aruna Olasekan Adekiya et al.,[1] A study examined how different organic manures and NPK fertilizers affected the quality, yield, and growth of okra (*Abelmoschus esculentus* L.) as well as the qualities of the soil. The study, which took place between 2017 and 2018, examined several organic manures, including pig, rabbit, cow, poultry, and green (Mexican sunflower) dung, in addition to NPK fertilizer. A control group received neither inorganic fertilizer or manure. The results demonstrated that by raising the amounts of organic matter, nitrogen, phosphorus, potassium, calcium, and magnesium in the soil, organic manures and NPK fertilizer greatly increased soil fertility. The beneficial effects of poultry manure on soil characteristics and okra productivity and growth stood out in particular. Higher protein, carbohydrate, mucilage, and mineral concentrations were among the quality features, growth, and yield of okra produced with poultry manure.

Shakib Mahmud Dipto et al.,[2] The study includes a review of the literature on various crop

management techniques and agricultural technologies, as well as a crop suggesting system based on N.P.K. values and machine learning models. Projects include classification techniques for crop prediction specific to soil, big data analysis for yield enhancement, and an Android app for crop recommendation. Furthermore investigated were agricultural simulation models, frameworks for predicting crop yield, and soil assessment using smartphones. Based on soil nutrient analysis, a fertilizer suggesting model was presented in a book published in Bangladesh by the Soil and Research Development Institute. The construction of the suggested system is influenced by these works together, which helps with agricultural practice decision-making.

Marianah Masrie et al.,[3] In order to monitor soil nutrients, the research article presents an optical transducer that capitalizes on the sensitivity and quick reaction of earlier techniques. The distinctive optical properties of potassium, phosphorus, and nitrogen in soil are highlighted, and a photodiode and LEDs are integrated for detection. Experiments assess soil absorption and optimize the transducer configuration. The findings classify nutrient levels by revealing information on photodiode responses and voltage readings for various soil samples. The successful creation of the transducer and its potential to overcome issues in soil nutrient determination are highlighted in the study's conclusion, along with suggestions for future research areas.

Shubham Prabhu et al.,[4] research article "Soil Analysis and Crop Prediction" focuses on creating a prediction

engine to identify the best crop for particular soil types. It highlights how important soil analysis is for determining crop growth variables like nitrogen, phosphorus, potassium, and hydrogen ion concentration as well as fertility. The paper proposes an automated soil testing system with the goal of delivering timely and affordable agricultural information. The process includes monitoring environmental conditions, analyzing soil samples from Mumbai, and creating an intuitive application that interfaces with the Internet of Things to transfer data in real time. The precision of temperature and soil moisture analysis with Arduino and cloud computing is demonstrated by the results, which help farmers manage their resources and increase agricultural productivity.

Vishal Kumar et al.,[5] A soil analysis system is suggested in the research article "Agriculture Soil Analysis for Suitable Crop Prediction" to help farmers choose the best crops for their areas. Optimizing crop yield requires soil testing, but the conventional approaches take a lot of time. Two datasets are used by the proposed system: crop data and soil imagery (red, black, mountain, and desert soil). Convolutional Neural Networks (CNNs) are used to anticipate appropriate crops and identify different types of soil. With precise crop suggestions based on soil type, the system seeks to raise agricultural production and farmer income. CNN techniques are used for model training in the system architecture, which uses digital image processing to determine the type of soil. The system's capacity to recommend crops based on soil categorization is highlighted in the conclusion.

Sita Rani et al., [6] The study presents a brand-new machine learning (ML)-based method that takes soil factors and weather into account when choosing crops for smart agriculture. In order to secure higher yields, it highlights how important it is to precisely select crops depending on agro-climatic conditions. The suggested model uses Random Forest Classifier for crop selection and LSTM RNN for weather prediction, and it achieves notably high accuracy in both domains. Compared to ANN, LSTM RNN performs better, showing lower RMSE for rainfall, maximum temperature, and

minimum temperature prediction. When it comes to crop selection, resource dependency prediction, and recommending the best times to plant, the Random Forest Classifier performs admirably. The Random Forest Classifier's model construction time is notably disclosed, which improves the methodology's transparency.

Dhruvi Gosai et al.,[7] research article "Crop Recommendation System using Machine Learning" explores an intelligent system that makes crop recommendations based on a variety of agricultural data. It examines previous research and comparative studies,

as well as supervised machine learning techniques for digital farming's effective crop recommendation systems and crop yield prediction. It also looks into AgroConsultant, a clever crop recommendation system that makes use of a variety of machine learning techniques. The usefulness of algorithms like Decision Tree, KNN, and SVM is highlighted in the paper's comparative examination of various systems. It also describes the technique of the suggested system, including data collection, preprocessing, and the use of algorithms such as Random Forest, Decision Tree, and Naïve Bayes. By improving crop recommendation systems through the use of machine learning algorithms, this research advances precision agriculture.

Maaz patel et al.,[8] According to the literature review, crop recommendation systems that use machine learning algorithms have a major emphasis on employing these algorithms to increase agricultural productivity and sustainability. Numerous methods have been investigated, such as neural networks, SVM, decision trees, and ensemble algorithms. Research has indicated that examining soil properties and utilizing algorithms such as random forest, CHAID, K-Nearest Neighbour, and Naïve Bayes can effectively optimize crop output. Using machine learning methods like decision trees, K-NN, random forests, and neural networks, intelligent systems like AgroConsultant have been built to forecast appropriate crops and rainfall. Future research should focus on scalability enhancements and real-time data integration to facilitate the broader implementation of these systems in agriculture and promote higher yields and sustainable agricultural methods.

Ajay Kumar Sahu<sup>1</sup> et al.,[9] The revolutionary role of virtual voice assistants—advanced AI-driven bots capable of comprehending human speech and carrying out tasks—is examined in this study. It explores how AI is changing the way people live their lives, emphasizing how user experiences are always being improved. Previous research indicates a significant increase in the use of virtual assistants, and positive developments are predicted for the future. Research highlights the importance of elements like trust and the quality of the information in influencing user adoption and acceptability. To illustrate useful applications, methodologies used in the construction of virtual assistants—most notably, Natural Language Processing algorithms and Python libraries—are reviewed. Use examples highlight the virtual assistants' extraordinary adaptability, especially when it comes to making them accessible to a wide range of user groups. Features that highlight both user enjoyment and economic efficiency are explained, and conversations about user preferences offer insightful information.

A.Rehash Rushmi Pavitra et al.,[10] The review of the literature delves into important topics that are necessary to create multilingual intelligent voice assistants. It talks about Multilingual Speech Recognition (MSR), covering methods and difficulties such as code-switching. It discusses strategies for Multilingual Speech Synthesis (MSS) and related difficulties. It also looks at Multilingual Voice Assistants (MVA), emphasizing well-known MVAs and difficulties with language modeling. Along with discussing obstacles and solutions, the survey also looks at developments in Multilingual Neural Machine Translation (NMT) and Voice Assistant Applications in Education. The fundamentals of Automatic Speech Recognition (ASR), which are essential for AI-based voice assistants, are finally explained. Together, these research improve our knowledge of and ability to use multilingual voice assistant technology effectively.

Pooja Darda et al.,[11] The Voice Assistant (VA) systematic literature review offers a thorough summary of research trends in this quickly developing subject. After providing a brief historical overview, the paper emphasises important contributions made by Azvine et al. (2000) and

Pokojski (2004) to the knowledge of VA software principles. It examines important aspects of commercial VA systems, like accessibility and naturalness, and references studies by Moore (2016) and Lopez, Quesada, and Guerrero (2018). Based on studies by Cowan et al. (2017) and Dubiel et al. (2018), opportunities for VA applications across sectors such as education and healthcare are identified, and existing limitations, such as privacy concerns and speech recognition accuracy, are discussed. The review's conclusion emphasises the necessity of filling in knowledge gaps on user expectations and advancing VA technology for usage in new applications.

Vishal Kumar Dhanraj et al.,[12] This research paper explores the process of creating a Python-based desktop voice assistant specifically tailored for Windows. The writers use the rich libraries and installer packages of Python, drawing inspiration from popular virtual personal assistants such as Cortana and Siri, to facilitate API calls, backend processing, and speech recognition. The research emphasises the importance of Natural Language Processing (NLP) and describes a locally hosted system architecture with the goal of enhancing security and privacy. The authors hope to further AI-driven human-computer interaction by offering insightful guidance for the continued development of their voice assistant through a thorough analysis of relevant publications that cover developing trends and HCI principles.

## **7. Conclusion**

An innovative agricultural system that uses automated Ultimately, the suggested approach makes use of machine learning algorithms and sensor technology to facilitate data-driven agricultural decision-making.

The device determines which crop will do best in a given situation by interpreting sensor readings from mud samples and evaluating critical factors including temperature, humidity, NPK levels, and soil moisture. Users can engage with the system more easily and comprehend it better because of the user-friendly interface and voice support available in both Kannada and English. In the end, this method maximizes agricultural output and promotes sustainable farming practices by enabling farmers to make knowledgeable crop selection

decisions. Farmer productivity and financial results can be enhanced while encouraging environmental conservation in contemporary agriculture by utilizing the power of cutting-edge technology.

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