

Precision Agriculture using ML for Soil and Weather Prediction

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Abstract: Agriculture is the backbone of many economies and plays a critical role in global food security. However, with growing challenges posed by climate change, erratic weather patterns, and soil degradation, the need for precise and predictive techniques in agriculture has never been more urgent. This paper focuses on the application of machine learning (ML) techniques in soil fertility prediction and weather forecasting, two critical components that can significantly impact agricultural productivity. By analysing soil properties, such as moisture, pH, and nutrient levels, and combining them with accurate weather predictions, our system aims to help farmers make informed decisions regarding crop selection, irrigation scheduling, and pest control. Leveraging algorithms like Naive Bayes for soil classification and Long Short-Term Memory (LSTM) networks for weather prediction, we provide a comprehensive solution for precision agriculture. This system not only enhances productivity but also promotes sustainable agricultural practices by optimizing resource use and reducing wastage.

Keywords: Machine Learning, Precision Agriculture, Soil Prediction, Weather Prediction, Crop Yield, Sustainable Agriculture

I. INTRODUCTION

Agriculture, being one of the oldest and most essential industries, is facing unprecedented challenges in the modern era. Rapid population growth, depleting natural resources, and climate change have significantly affected the agricultural sector. Traditional farming methods, which rely on experience and historical knowledge, are often inadequate to address these challenges in today's dynamic environment. There is a growing need for more accurate, data-driven approaches to agriculture, especially when it comes to predicting soil health and weather conditions.

Precision agriculture aims to increase crop yield and resource efficiency by utilizing advanced technologies such as machine learning (ML) and data analytics. By predicting soil fertility and weather patterns, farmers can make more informed decisions that improve productivity while minimizing environmental impact. The primary goal of this research is to develop a system that integrates soil prediction and weather forecasting using machine learning algorithms, offering farmers a reliable tool for optimizing their farming practices.

II. LITERATURE SUREVY

Numerous studies have explored the potential of machine learning in agriculture. Fatin Farhan Haque et al. (2021) used Support Vector Regression (SVR) and Linear Regression (LR) to predict crop yields based on various environmental factors such as UV exposure, water usage, and soil nutrient levels. Their model achieved an impressive R^2 score of 0.85, demonstrating the effectiveness of ML in crop yield prediction. Ramesh Medar (2021) developed a Random Forest-based system for predicting crop yields by analysing historical data, such as rainfall, temperature, and soil quality, showing significant improvements in crop selection efficiency.

In contrast to traditional methods of crop planning, machine learning allows for dynamic analysis of real-time data, leading to more accurate predictions. Similarly, Prof. A.V. Deorankar (2021) highlighted the use of image processing for soil and land classification, demonstrating how remote sensing technologies can enhance soil analysis by classifying soil health and suggesting optimal crops. Despite these advances, there remains a gap in the integration of soil fertility

and weather prediction models into a single system. Our research addresses this gap by combining soil and weather predictions for a more comprehensive precision agriculture solution.

III. EXISTING SYSTEM

Current agricultural management systems rely on fragmented data sources, often limited to either soil analysis or weather forecasts. For instance, platforms like Meteo Group focus primarily on weather prediction but fail to incorporate soil characteristics, which are equally important in determining the success of agricultural operations. On the other hand, soil testing methods are typically manual, time-consuming, and region-specific, making it difficult for farmers to access timely and accurate information on soil fertility.

Other systems, such as Climate FieldView, provide basic weather insights and soil monitoring but are often constrained by the lack of real-time adaptability and region-specific customization. These limitations result in poor predictions and hinder the adoption of precision farming techniques at scale. Our proposed system addresses these shortcomings by integrating soil prediction and weather forecasting in a unified, machine learning-powered platform. The result is a comprehensive solution that empowers farmers with accurate, timely, and actionable data.

IV. PROPOSED SYSTEM

The proposed system integrates machine learning techniques to predict soil fertility and weather conditions, delivering a holistic precision agriculture solution. The system is composed of two key modules:

1. Soil Prediction Module: This module leverages machine learning classifiers such as Naive Bayes and Random Forest to predict soil fertility based on critical parameters like pH, nitrogen, potassium, and phosphorus levels. Random Forest is particularly effective in handling large datasets and improving accuracy. The module also considers geographical attributes and climate history to recommend suitable crops for specific soil types. The soil prediction process follows several key steps:

- Data Collection: Soil data is sourced from agricultural databases and surveys, while additional data on soil moisture and temperature is collected from sensors placed in the field.
- Feature Extraction: Soil attributes such as texture, organic matter content, and moisture levels are extracted from the dataset.
- Model Training: A Random Forest classifier is trained on this data, with cross-validation used to improve model accuracy.
- Prediction Output: The system predicts soil fertility levels and suggests optimal crops based on the local soil profile.

2. Weather Prediction Module: The weather prediction module utilizes Long Short-Term Memory (LSTM) networks, which are well-suited for time-series forecasting. LSTM networks can model complex temporal patterns in climate data, such as rainfall, humidity, and temperature. These predictions help farmers determine irrigation schedules, manage water resources, and plan planting cycles. The weather prediction process involves:

- Data Collection: Weather data, including historical and real-time meteorological information, is sourced from local weather stations and satellite networks.
- Modelling: LSTM networks are trained on this data to capture seasonal trends and predict future weather conditions.
- Prediction Output: The model forecasts critical weather parameters such as rainfall, temperature, and humidity, helping farmers mitigate risks like droughts or excessive rainfall.

Together, these two modules provide a unified platform that delivers real-time recommendations on optimal planting schedules, crop choices, and irrigation management. This approach not only improves yield but also promotes resource conservation by minimizing water and fertilizer usage.

V. CONCLUSION

This project developed a comprehensive model for soil classification and crop prediction, focusing on identifying suitable crops based on soil properties and weather conditions. By using machine learning algorithms such as Naive

Bayes for soil classification and Random Forest for crop recommendation, the system achieved significant accuracy in predicting soil fertility and providing appropriate crop suggestions. The integration of these technologies simplifies decision-making for farmers, enabling them to maximize productivity and minimize resource wastage.

Additionally, the model's predictive capability for weather forecasting enhances its utility by helping farmers plan their irrigation and planting schedules effectively. The combined soil and weather prediction system offers a robust solution for modern agriculture, addressing issues such as climate unpredictability and soil degradation. Future work can focus on expanding the model to include real-time sensor data and broader geographical coverage to further enhance its effectiveness.

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