

Land Insight Platform Management System using Machine learning

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Abstract—The AI-driven platform supports real-time monitoring and decision-making processes, thereby optimizing crop yields and minimizing resource wastage. It leverages machine learning algorithms to predict the most suitable fruit crops for specific soil types and weather conditions, ensuring sustainable agriculture practices. The platform also facilitates farmer education and training through its user-friendly interface, providing insights and recommendations for optimal land use. Furthermore, it includes data analytics tools to track and assess long-term soil health and productivity. The case study highlights significant economic benefits for farmers, including increased profitability and reduced input costs. The platform's design combines key technologies, daily business requirements, and system design principles analysis to ensure efficient land resource allocation. By fostering a data-driven approach to land management, the platform paves the way for a more resilient and sustainable agricultural sector.

Index Terms—Land Resource Management, Artificial Intelligence, Information Platform, Machine Learning, Naive Bayes, Support Vector Machine (SVM), System Design, Data Management.

I. INTRODUCTION

Land resource planning and management are crucial for national economic development and sustainable land use. Effective planning optimizes land value and ensures the harmonious development of regional economy, society, and ecology. Current land management practices often rely on manual processes, which are inefficient and prone to errors. These methods struggle to handle the large volumes of data

involved in land resource management and lack integration with modern technological advancements.

The rapid development of information technology, particularly artificial intelligence (AI), cloud computing, and big data, offers new solutions for improving land management. Integrating AI into information management platforms can provide faster, more efficient, and accurate methods for data query, analysis, and decision-making. This paper outlines the development of a Land Resource Management Information Platform based on AI technology. The goal is to leverage AI to enhance the efficiency and accuracy of land resource management and planning, assisting land departments in comprehensive supervision and improving business informatization. In this introductory section and we outline the motivation behind our research and provide an overview of existing methodologies and highlight the significance of our proposed CNN based approach. We anticipate that our findings will contribute to the ongoing discourse in the field of image recognition and providing valuable insights for researchers and practitioners working on handwritten character recognition and paving the way for enhanced solutions in real world applications.

This paper introduces the recognition of handwritten digits (0 to 9) from the well-known MNIST dataset using the TensorFlow framework (library) and Python as the programming language, along with its associated libraries. Upon inputting a specific digit, the system is designed to recognize and display the

results with a high level of accuracy.

II. DATA SOURCE AND STATEMENT

Conventional land management information systems in China have primarily been designed to oversee the administration of land assets and resources. These systems rely heavily on manual processes, utilizing data collected from field mapping, surveys, and historical records. While functional to a certain extent, such approaches often result in inefficiencies and introduce potential inaccuracies in planning and decision-making. A significant drawback is the absence of intelligent or autonomous capabilities, which limits the adaptability and relevance of these systems in the context of contemporary and future socio-economic development.

In light of these limitations, there is an evident necessity for the development and deployment of an automated, intelligent land management system. Such a system should be capable of processing large volumes of data, conducting advanced analytical operations, and supporting concurrent multi-user access—demonstrated to be scalable up to 200 users in the proposed framework. Furthermore, it must serve as a reliable decision support tool for land planning authorities.

The data handled by this intelligent system encompasses various land resource parameters, including but not limited to soil composition (percentages of sand, silt, and clay), moisture levels, pH values, mineral content, organic matter concentration, and the presence of air and water within the soil. These variables serve as critical input features for predictive models, enabling more informed and precise land-use planning decisions

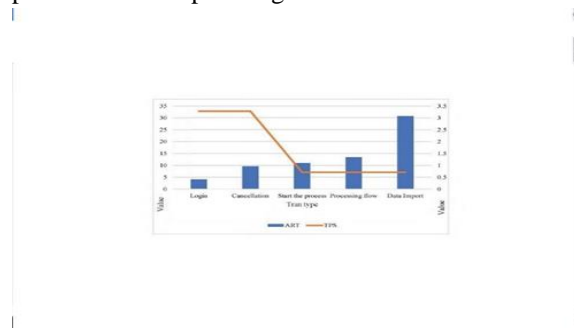


Fig1 System Design

III. PROPOSED SYSTEM AND METHODOLOGY

A. Pre Processing

Effective data preprocessing is crucial. Techniques employed include removing noise, handling missing information, modifying default values, and potentially grouping attributes for prediction. The dataset used for machine learning requires features that capture the characteristics of land objects.

We have converted our image from RGB to Gray scale for easy computation by dividing with 255-pixel value to get the value between 0-1. In this process we are also converting the data from categorical

B. System Architecture and Core Technologies

The platform is structured into four key modules: User, Admin, Data Preprocessing, and Machine Learning, each serving a distinct purpose to ensure smooth and effective system functionality. The User Module allows end-users to interact with the platform, enabling actions such as registration, login, data input, and accessing prediction results. The Admin Module provides administrative control over user access, data management, and system monitoring. The Data Preprocessing Module handles the critical tasks of cleaning, normalizing, and transforming raw land resource data, ensuring that it is properly structured and formatted for analysis.

The Machine Learning Module is responsible for executing predictive algorithms and generating analytical insights that support decision-making processes related to land use and classification.

From a technological standpoint, the platform's backend is developed using Python with the Django framework, selected for its robust structure, security features, and scalability for web application development. Data is stored using SQLite, a lightweight relational database system well-suited for handling structured datasets in a streamlined environment. The system integrates multiple machine learning algorithms, including Naive Bayes, Support Vector Machine (SVM), and Decision Tree classifiers. Naive Bayes and SVM are chosen due to their effectiveness in classification scenarios—commonly applied in sentiment analysis and adapted here to evaluate land resource parameters—while the Decision Tree classifier is employed for its interpretability and effectiveness in rule-based decision-making tasks. This combination of technologies and methodologies ensures the platform

delivers accurate, reliable, and user-friendly predictions.

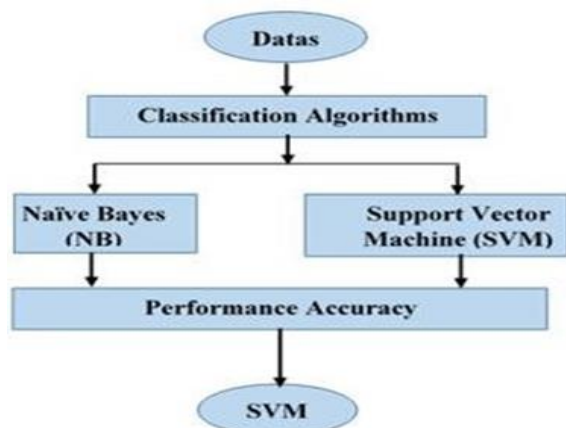


Fig 2 Naïve Bayes and SVM

C. Functionality

The platform features a user-friendly User Interface* that supports essential functionalities such as user registration with input validation, login access following admin approval, data uploading and viewing, as well as initiating land suitability predictions based on provided parameters. Complementing this is the Admin Interface, which enables administrators to manage user access by activating newly registered users, oversee system-wide data, and monitor the performance of machine learning algorithms through key evaluation metrics, including accuracy, precision, recall, and F1-score.

The Data Handling and Analysis component of the system encompasses comprehensive data preprocessing procedures, ensuring that the dataset is cleaned and prepared for analysis.

The data is then split into training and testing sets in a 60:40 ratio, enabling users to perform classification tasks and generate predictions with meaningful insights given from the machine learning models.

IV. SYSTEM TESTING AND RESULTS

To ensure the reliability, stability, and correctness of the platform, comprehensive testing was conducted across all components and modules. The testing process incorporated multiple types of evaluation, including Unit Testing, which validated the internal logic of individual functions; Integration Testing, which assessed the interaction between interconnected

components; Functional Testing, which verified each feature's behavior against the specified requirements; and System Testing, which evaluated the performance and correctness of the fully integrated system. Both White Box (code-level logic validation) and Black Box (input/output-based testing without internal code awareness) methodologies were employed to cover all possible error conditions and edge cases.

Classification of Recognition results



Fig 3 Support Vector machine

The testing strategy focused on verifying input field validations, proper activation of user interface links, system response times, and preventing duplicate data entries. A variety of sample test cases were executed, including scenarios related to user registration, login functionality, data upload and viewing, machine learning model execution, prediction accuracy, and administrative tasks such as user activation.

The results of the testing phase were highly satisfactory. All test cases passed without identifying any defects. Both functional and performance evaluations confirmed that the system meets its intended design specifications and operates efficiently under expected workloads. The platform successfully demonstrated support for concurrent multi-user operations, validating its scalability and robustness. Visual evidence of the system's successful test execution is provided through annotated screenshots included in the original documentation.

V. PREDICTION CAPABILITIES

One of the core functionalities of the platform is its robust predictive capability, which enables users to evaluate land suitability based on specific input parameters. Users are prompted to enter various land characteristic data such as the percentages of Sand,

Silt, and Clay, the Dry or Moist state, pH level, Mineral content, Organic Matter, as well as Air and Water content.

Once this data is submitted, the system processes it using a pre-trained machine learning model—specifically, a Decision Tree classifier as implemented in the current sample configuration. Based on the analysis, the model predicts the land's suitability classification, which may include categories such as Desert, Forest, Suitable for Agriculture or Construction, or Not Suitable for Cultivation. This functionality empowers users, particularly land planners and stakeholders, to make informed decisions about land use by leveraging data-driven insights generated through the platform's intelligent analytics.

The screenshot shows a web application titled "Machine Learning". It has a navigation bar with links: Home, ViewData, PredictLand, NaiveBayes, SVM, and Logout. The main section is titled "Add Information to Test". Below this, there is a table with two columns: "Fields" and "Input Values". The rows are:

- Sand(%) in Land: [Text Input]
- Silt(%) in Land: [Text Input]
- Clay(%) in Land: [Text Input]
- Dry: [-Select-]
- Moisture: [-Select-]
- pH: [Text Input]
- Minerals: [-Select-]
- Organic Matter: [-Select-]
- Air in Land: [-Select-]
- Water: [-Select-]

 At the bottom of the form is a "Predict" button.

Fig 4 Predicting Land

V. CONCLUSION AND FUTURE ENHANCEMENT

This paper presented a comprehensive overview of the analysis, design, development, and testing of an AI-driven Land Resource Management Information Platform. The system effectively addresses the limitations of traditional, manual land management approaches by integrating artificial intelligence techniques, particularly Naive Bayes and Support Vector Machine (SVM) algorithms. Through structured modules and user-friendly interfaces, the platform enables streamlined data management, predictive analysis, and informed decision-making. Functional and performance testing confirmed that the platform meets its intended objectives, demonstrating reliability, scalability, and accuracy in multi-user environments.

The platform is built using a modular architecture and features intuitive, user-friendly interfaces that support a variety of key functions. These include efficient data management, predictive analytics, and decision support tools. The incorporation of AI enables automated classification of land types, forecasting of land use trends, and the provision of insights to guide policy-making and resource allocation. This makes the system highly valuable for stakeholders such as government agencies, urban planners, environmental managers, and agricultural developers.

Looking forward, several future enhancements can be considered to further strengthen the platform's capabilities. These include optimizing the performance and accuracy of the AI models, incorporating larger and more diverse datasets to improve generalizability, and enabling real-time data integration for dynamic updates. Additionally, the system could benefit from enhanced Geographic Information System (GIS) features to support advanced spatial analysis and visualization. Exploration of more sophisticated AI architectures, such as ensemble models or deep learning frameworks, may also provide improved predictive performance and adaptability across varied land resource scenarios.

VI. REFERENCES

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