SMART FARMING SOLUTION

UIT2602 – WEB PROGRAMMING

A PROJECT REPORT

Submitted by

Sriram M- 3122 22 5002 136 Tharun Kumar S- 3122 22 5002 147

SSN COLLEGE OF ENGINEERING, KALAVAKKAM

APRIL 2025

Sri Sivasubramaniya Nadar College of Engineering (An Autonomous Institution, Affiliated to Anna University)

BONAFIDE CERTIFICATE

Certified that this project titled "Smart Farming Solutions" is the
Bonafide work of "Sriram M–3122 22 5002 136, Tharunkumar S–3122
22 5002 147 and is submitted for project End Semester Examination held
on <date>.</date>

Signature of examiner

Submitted on -----

Internal Examiner

External Examiner

1. ABSTRACT:

Smart farming leverages modern technologies to enhance productivity and decision-making in agriculture. This project presents a comprehensive smart farming solution that provides real-time crop prices, crop and fertilizer recommendations, and data-driven insights to assist farmers and stakeholders.

By integrating machine learning models such as Random Forest Regressor and Decision Trees, the system forecasts crop prices and suggests optimal crops and fertilizers based on soil and environmental conditions.

The platform includes both a web interface and a mobile application developed using Flutter, connected to a Flask-based backend via Node.js APIs, ensuring accessible and user-friendly interaction for diverse users. Additionally, an inventory management system has been integrated, utilizing MongoDB Atlas for efficient cloud-based data storage and performing full CRUD operations. This system manages farm products, equipment, sales details, customer information, and also supports sales reporting with export to Excel functionality, offering streamlined and organized resource tracking for better farm management.

2. INTRODUCTION:

Agriculture remains vital to global food security, yet traditional methods—driven by experience and intuition—can lead to inefficiencies and missed market opportunities. This Smart Farming Solution harnesses machine learning, cloud computing, and cross-platform development to deliver real-time insights and streamline farm operations.

Core Features

- **Real-Time Crop Pricing:** Live tracking and visualization for nearly 100 crops to guide selling decisions.
- Intelligent Recommendations:
 - Crop Selection: Suggests best crops based on soil (pH, NPK)

- levels) and environmental factors (temperature, humidity, rainfall).
- Fertilizer Use: Recommends optimal fertilizer mixes to boost yield and reduce waste.
- **Price Forecasting:** Uses Random Forest and Decision Tree models to predict price trends, presented in clear graphical form.
- **Multi-Platform Access:** Responsive web portal (Flask + Node.js APIs) and a Flutter mobile app ensure farmers have insights at their fingertips.
- **Inventory Management:** MongoDB Atlas-backed module with full CRUD for products, equipment, customers, and sales—plus Excel export for reporting.

By combining data-driven analytics with an intuitive interface, this platform empowers farmers to optimize resource use, increase profitability, and adapt quickly to market dynamics—all from one unified system.

3. LITERATURE SURVEY:

- A range of tools and studies underpin the components of our Smart Farming Solution, though few combine them end-to-end.
- Market Price Tracking:

Platforms like AgriPrice (2018) and CropSense (2020) aggregate real-time commodity prices and offer trend dashboards or SMS alerts, but typically cover a limited crop set.

Recommendation Systems:

ML-based models such as CropAdvisor (2019) and AgriSmart (2021) use Random Forests and decision trees

on soil and weather data to boost yield predictions and optimize rotations, yet rarely integrate fertilizer guidance alongside crop choice.

Price Forecasting:

Tools like PricePredict (2020) and AgriCast (2023) employ ARIMA, LSTM, and ensemble regressors to forecast prices with under 10–18% error, but often lack seamless linkage to planting or sales modules.

Inventory Management:

Solutions such as FarmTrack (2020) and AgriERP (2022) provide cloud-based CRUD for farm inputs, equipment, and sales with reporting features (including Excel export). However, they operate as standalone ERPs without real-time analytics or recommendations.

4. DELIVERABLES:

Machine Learning Models

- Random Forest Regressor for real-time crop price forecasting
- Decision Tree Classifier for crop and fertilizer recommendation

Mobile Application (Flutter)

- Real-time data visualization via RESTful API calls (crop prices, forecasts)
- Interactive crop and fertilizer recommendation screens
- Inventory management UI for CRUD operations on products, equipment, customers, and sales
- Sales report generation and export to Excel

Web Dashboard (Node.js / Express)

- Crop Price Forecast: graphical view of predicted price trends
- Crop Recommendation: input-driven suggestions based on soil & environmental data
- Fertilizer Recommendation: tailored nutrient plans for selected crops
- Inventory Management: full CRUD interfaces for farm products, equipment, customer records, and sales transactions
- Sales Reporting: tabular sales summaries with Excel export functionality

Backend Services

- Flask microservices hosting ML endpoints
- Node.js / Express APIs for data retrieval, recommendation logic, and inventory operations
- MongoDB Atlas cloud database for scalable, secure storage of all application data

5. Requirement Engineering

Sprint #	Epic	User Story #	Requirement / User Story	Essential or Desirable	Description of the Requirement	Remarks
	Crop Price Data Acquisition	1	Develop a user-friendly interface	Essential	Develop a module for acquiring crop price data from appropriate datasets.	Ensure variety and correctness of the data
2	Data Preprocessing	2	Artifact removal And Feature Extraction	Essential	Develop methods for removing Artifacts like unused indices and extract features	Use appropriate algorithms/ techniques for artifact removal and extract features.
3	Random Forest Training	3	Train Random Forest model.	Essential	Train an Random Forest model using preprocessed crop price data	Evaluate the model using evaluation techniques.
4	Random Forest, Decision Tree, XgBoost, Logistic Regression Training	4	Train Random Forest, Decision Tree, XgBoost, Logistic Regression	Essential	Train Random Forest, Decision Tree, XgBoost, Logistic Regression using crop recommendation data	Evaluate the model using evaluation techniques.

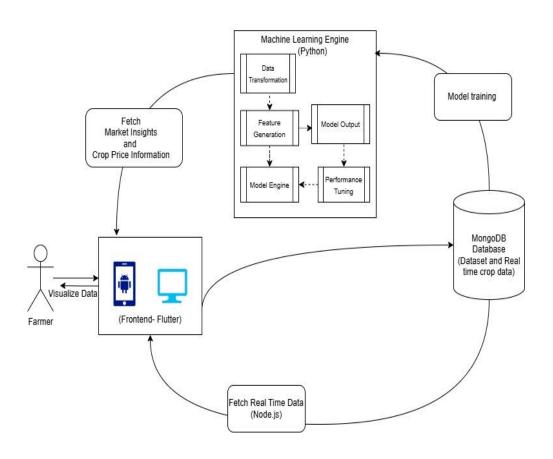
5	Random Forest, Decision Tree, XgBoost, Training	4	Train Random Forest, Decision Tree, XgBoost	Essential	Train Random Forest, Decision Tree, XgBoost, using fertilizer recommendati on data	1
---	---	---	---	-----------	--	---

6. RISK MANAGEMENT

Risk#	Risk Description	Probabilit y	Impact	Mitigation Plan
Consistency Issues	Inconsistent normalization across different subjects could lead to unreliable model training and evaluation.	Medium	High	Implement standard procedures for normalization and ensure they are uniformly applied to all datasets.
Model Overfitting	The Transformer model might overfit to the training data, reducing its performance on unseen data.	Medium	High	Use regularization techniques such as dropout. Implement cross- validation and monitor performance on a validation set.
Training Time	Training Transformer models can be very time-consuming, especially with large crop price. Crop recommendation, fertilizer recommendation datasets.	Medium	Medium	Utilize powerful hardware (GPUs/TPUs). Optimize training process with techniques like batch normalization and early stopping.

7. ARCHITECTURE DIAGRAM:

SMART FARMING SOLUTIONS



The architecture of the smart farming solution is built on a modular and scalable design that combines machine learning intelligence, a robust backend, and interactive frontends across web and mobile platforms. At the heart of the system lies the machine learning layer, which powers three primary functionalities: crop price forecasting, crop recommendation, and fertilizer recommendation.

The crop price forecasting model utilizes a Random Forest Regressor trained on historical market data for over 100 crops. This model takes in various features such as crop type, region, seasonality, and historical trends to predict future prices. It outputs insights like the top five crops expected to gain or lose in price and provides a 12-month forecast visualized through graphs. Similarly, the crop recommendation system, built using a Random Forest

Classifier, processes environmental and soil parameters like pH, temperature, rainfall, and humidity to recommend the most suitable crops. For fertilizer recommendation, a Decision Tree model is employed, which takes input such as soil nutrient values (NPK) and crop type, then suggests the appropriate fertilizer type and dosage.

These machine learning models are trained offline, saved as serialized .pkl files, and loaded into a Flask-based backend. The Flask application serves as the API layer, handling all incoming requests, preprocessing data, running predictions using the loaded models, and returning results in JSON format. The backend exposes endpoints for each functionality—price prediction, crop suggestion, and fertilizer advice—making it easy for frontend interfaces to access the intelligence.

On the web frontend, a combination of HTML, CSS, and JavaScript is used to build a user-friendly interface. The web pages interact with the backend through API calls, which are optionally routed via a Node.js server. The web interface provides users with dashboards to view price trends, input forms for crop and fertilizer recommendations, and dynamic visualizations using tools like Chart.js or D3.js.

For users on mobile devices, a Flutter-based mobile application provides a seamless experience. This cross-platform app connects directly to the Flask APIs using HTTP requests and presents results through an intuitive interface. Features include real-time crop price dashboards, interactive input forms for recommendations, and forecast visualizations using packages like fl_chart. The app also supports basic offline caching for areas with limited connectivity.

Communication between the frontend and backend is handled via standard RESTful APIs, ensuring a clean and decoupled interaction. The backend is designed to be stateless, scalable, and ready for cloud deployment on platforms like Heroku, Python Anywhere. While the current system operates without a central database, it can be extended with Firebase, SQLite, or MongoDB to store user history, preferences, or logs.

In summary, the architecture integrates intelligent machine learning models with a flexible backend and dual-platform frontend systems. This design ensures that farmers and agricultural stakeholders can access powerful insights, recommendations, and forecasts—anytime, anywhere, on both web and mobile platforms.

8. ML MODEL DESCRIPTION:

Crop Price Prediction

Model: Random Forest Regressor

Data: Historical market prices of 100+ crops from government and private

sources.

Features: Crop name, location, time, supply, demand metrics.

Output:

• Top 5 gainers (price increase)

• Top 5 losers (price drop)

Next 12-month price forecast with visual graphs

 Justification: Random Forest performs well in handling timeseries data with non-linearity and avoids overfitting.

Crop Recommendation

Model: Random Forest Classifier

• Inputs: Soil type, pH value, rainfall, temperature, humidity

Output: Best-suited crops for the given conditions

Fertilizer Recommendation

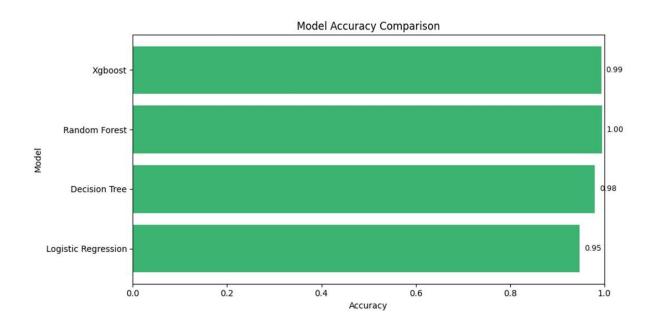
• Model: Decision Tree

• Inputs: Crop type, soil nutrient levels (N, P, K), pH

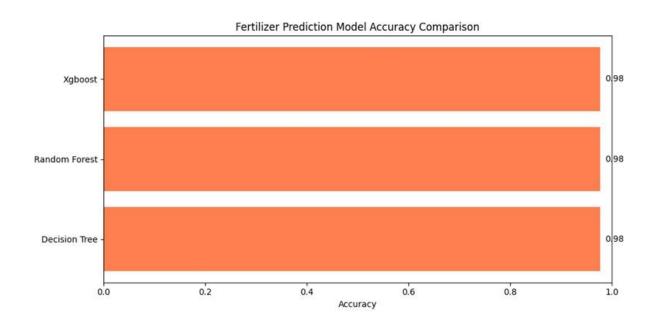
Output: Suitable fertilizer(s) and dosage

9. MODEL EVALUATION:

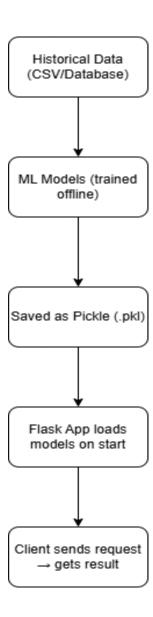
1.Crop Recommandation System



2.Fertilizer Recommandation System



OVERVIEW DIAGRAM:



10. MOBILE APPLICATION DESCRIPTION:

Platform: Flutter (cross-platform)

Connectivity: REST API calls to the Flask and Node.js backends

Features:

User Login & Registration

- Real-Time Crop Prices Dashboard
- Crop Recommendation Form with interactive result display
- Fertilizer Suggestion Module based on soil and crop data
- Forecast Graphs (e.g., via the fl_chart package)

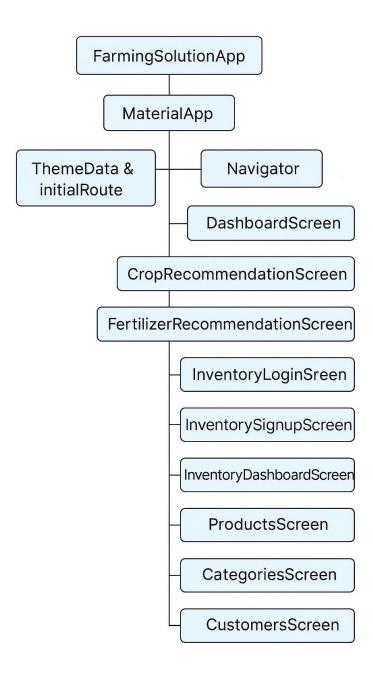
Inventory Management

- CRUD operations for farm products, equipment, customer profiles, and sales records
- View and update stock levels on the go
- Generate and export sales reports to Excel

UI/UX

- Simple, intuitive design optimized for rural connectivity
- Offline caching of last-fetched data (prices, recommendations, inventory) for continued access in low-network areas

11. MOBILE OVERVIEW WIDGET DIAGRAM:



12. WEB DESCRIPTION:

Frontend: HTML/CSS/JavaScript

Backend: Flask (Python) microservices, exposed via Node.js / Express

APIs

Pages:

Dashboard

Price "winners" and "losers" widgets
Interactive visualizations of current market data

Crop Recommendation

Input form for soil & environmental parameters Dynamic result display of suggested crops

Fertilizer Recommendation

Tool to select crop and input soil nutrients Output of optimal fertilizer mix

Price Forecast

Interactive charts showing monthly price trends and forecasts

Inventory Management

Products & Equipment: Create, read, update, delete entries **Customer Records:** CRUD customer profiles and contact info **Sales Module:** Log transactions, adjust stock levels

Reports: Generate sales summaries and download as Excel

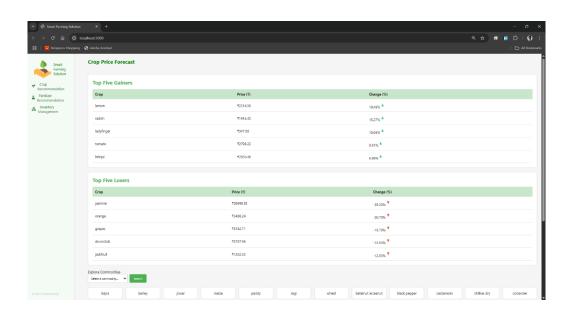
Tools & Libraries:

- Charts & Graphs: Chart.js and D3.js for interactive, responsive visualizations
- **Data Tables:** Ploty lightweight table library for inventory grids
- Styling: CSS Grid for layout, responsive design for desktop and tablet

13. LIMITATIONS AND CONSIDERATIONS:

These results indicate the model is ready for deployment as a highly accurate, cost-effective irrigation trigger that could reduce hardware complexity. However, the system requires further validation across seasonal variations and different environmental conditions to ensure robustness against sensor noise and changing agricultural patterns. Periodic model retraining and continuous performance monitoring are recommended to maintain this high level of accuracy in real-world applications, where factors like soil type variability and weather extremes may impact performance. The findings strongly support implementing a moisture-centric monitoring approach while highlighting the need for ongoing evaluation as the system scales to different farming conditions.

14. WEB INTERFACES:



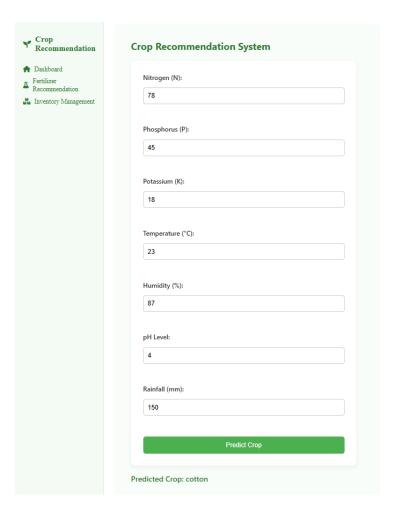


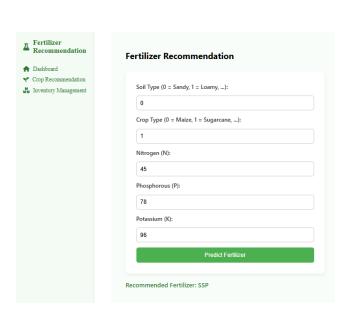
Forecast Trends		
Month	Price (per Qtl.)	Change
May 25	₹2871.88	-0.05% ▼
Jun 25	₹2873.34	0% 🔺
Jul 25	₹2890.54	0.6%
Aug 25	₹2890.54	0.6%
Sep 25	₹2873.34	0% 🔺
Oct 25	₹2871.88	-0.05% ▼
Nov 25	₹2873.3	0% 🔺
Dec 25	₹2863.94	-0.33% ▼
Jan 26	₹2872.74	-0.02% ▼
Feb 26	₹2898.74	0.89% 🔺
Mar 26	₹2863.94	-0.33% ▼
Apr 26	₹2873.3	0% 🔺

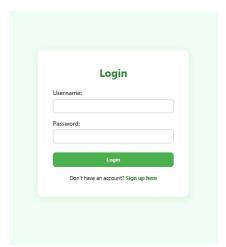


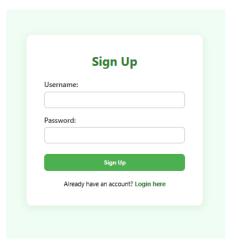
Previous Year Price Trend

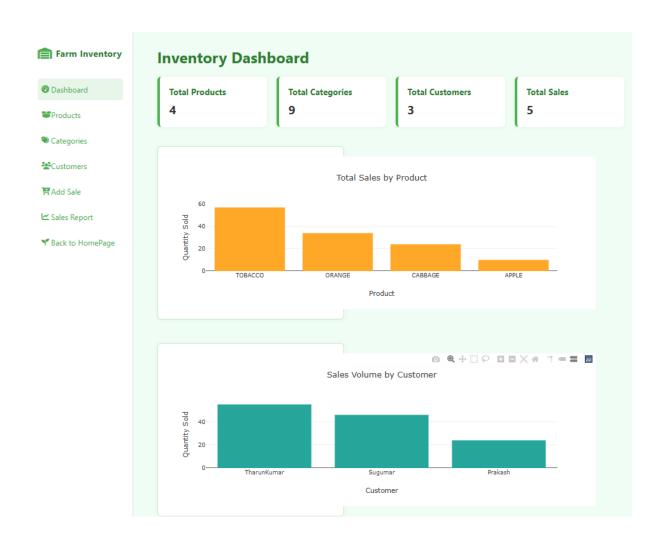


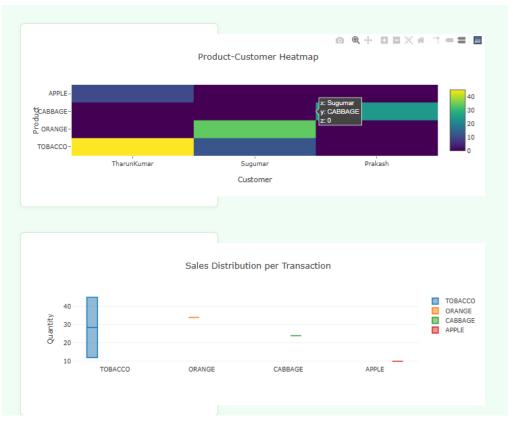


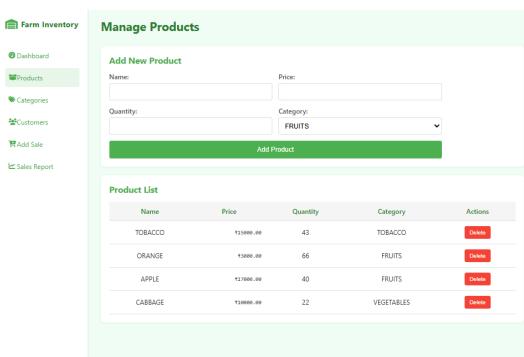


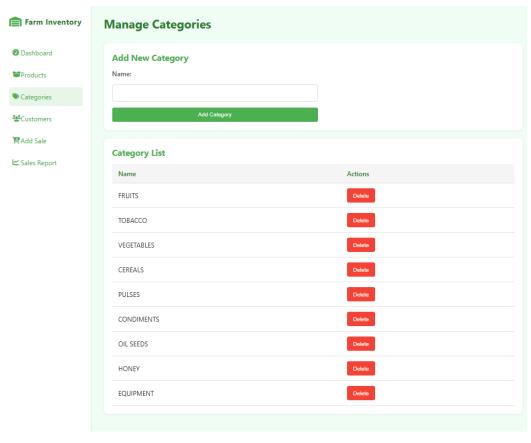


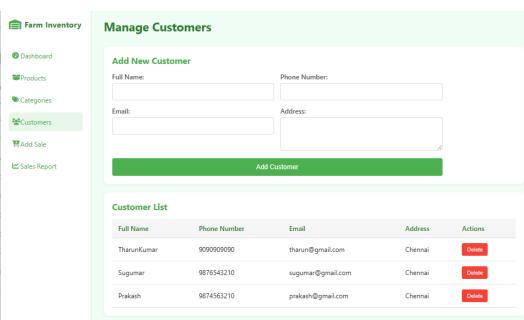


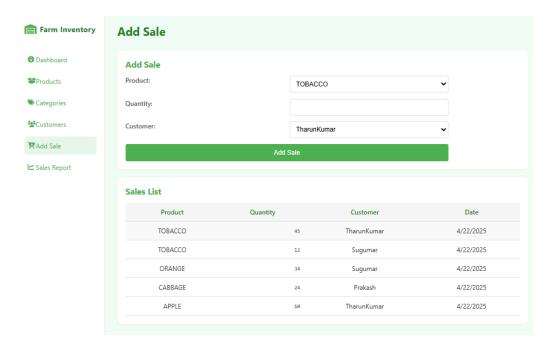


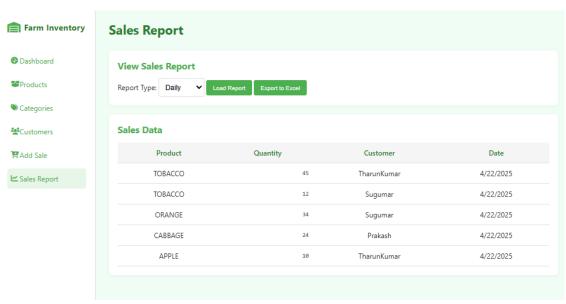




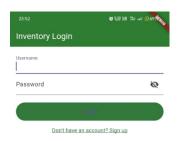




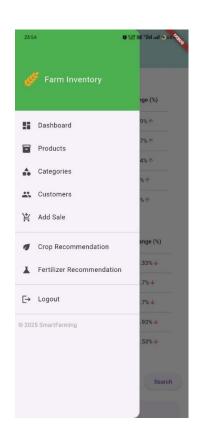




15. MOBILE APP INTERFACES:









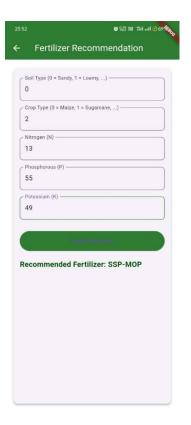






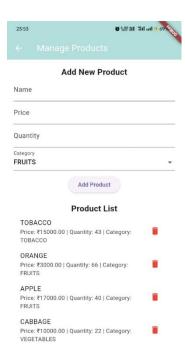


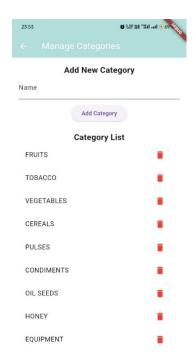


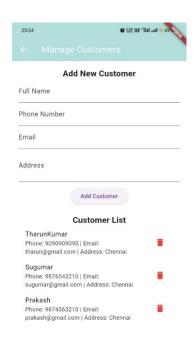


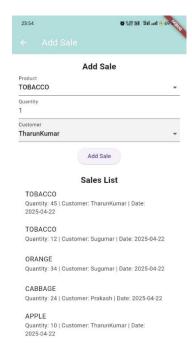












16. **FUTURE SCOPE:**

- IoT Integration: Add support for sensor data for real-time soil and weather monitoring.
- Multilingual Support: Enable regional languages in the UI for broader adoption.
- Blockchain Integration: For secure, transparent crop transactions and supply chain traceability.
- Community Forum: Add farmer forums or chatbots for peer advice.

17. CONCLUSION:

This Smart Farming Solution delivers an integrated, data-driven platform that empowers farmers across the entire production cycle—from market price awareness to crop selection, fertilizer optimization, and resource management. By harnessing machine learning models for price forecasting and recommendations, providing real-time dashboards, and embedding a MongoDB Atlas-backed inventory module (with full CRUD and Excel reporting), the system tackles critical challenges of yield maximization, input efficiency, and operational transparency. Although real-world deployment will require ongoing refinement such as expanding data sources, enhancing offline resilience, and fine-tuning ML accuracy this unified, tech-enabled framework establishes a robust foundation for sustainable and profitable agriculture.

18. REFERENCES:

- Wang, L., & Kumar, S. (2020). CropSense: SMS-based commodity price alerts for rural farmers. Journal of Agricultural Extension, 28(3), 157–165.
- Patel, A., Sharma, N., & Verma, S. (2018). AgriPrice: Automated web crawler system for real-time crop price aggregation. International Journal of Agricultural Information Systems, 9(2), 98–110.
- Li, J., Singh, R., & Mehta, V. (2017). NutriFarm: SVM-based fertilizer recommendation using soil nutrient profiles. Precision Agriculture, 18(4), 660–678.
- Flutter Documentation. (2025). Flutter: UI Toolkit for Building Natively Compiled Applications. Retrieved April 2025, from https://flutter.dev/