

Machine Learning Based Farming Assistant

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30-09-2024

Abstract

Agriculture is one of the vital industries of the country that forms the foundation of the global food supply, supporting the livelihoods of billions of people worldwide. The agricultural market is quite volatile due to certain reasons like climate vulnerability, knowledge gap, limited resources etc. which negatively affects the growth of the market. This report introduces “Farmologist”, an ultimate farming assistant that leverages Machine Learning or in the broader sense AI that aims to remove the voids in agricultural market. Given the vast agricultural market, Farmologist focuses on enhancing productivity and sustainability for small-scale agriculture and Organic farming.

The smart farming app will help farmers by providing real-time data and insights on weather, soil health, and crop conditions. It will use AI to offer personalized recommendations on when to plant, irrigate, and harvest, improving crop yields and reducing costs. The app will also facilitate access to market prices and financial services, including crop finance loans, helping farmers make informed decisions and secure better income.

1. Problem Statement

1. **Low Crop Yields:** Farmers struggle with low productivity due to inadequate access to timely and accurate agricultural information.
2. **Weather Uncertainty:** Unpredictable weather patterns lead to crop losses and financial instability.
3. **Lack of Market Access:** Farmers often sell their produce at low prices due to limited knowledge of market rates.
4. **Pest Infestations :** Large outbreaks of pests and plant diseases that damages crop without treatment.
5. **Limited Financial Support:** Difficulty in accessing crop finance loans and other financial services hampers their ability to invest in better farming practices.
6. **Resource Mismanagement:** Inefficient use of water, fertilizers, and pesticides due to lack of real-time data.
7. **Technological Barriers:** Rural farmers have limited access to advanced farming technologies and tools, widening the gap between traditional and modern farming practices.

Solving the problems faced by farmers is crucial as these issues directly impact overall food production, leading to reduced crop yields and higher food insecurity. Challenges such as poor soil health, water scarcity, and market access hinder farmers' ability to produce enough food efficiently resulting in financial instability of the farmers. AI and other technologies can address these problems by optimizing resource use, improving crop management, and streamlining market access, ultimately enhancing productivity and ensuring a stable food supply for growing populations.

2. Market/Customer/ Business Needs Evaluation

2.1 Market Analysis

The agricultural market is crucial in India as it significantly impacts the economy, contributing around 17-18% to the GDP and employing over 40% of the workforce. It ensures food security for a large population, drives rural development, and supports local economies through farmers' markets and trade. There are Commodity Markets that includes major crops like wheat, corn, rice, and soybeans, as well as livestock and dairy products and Farmer Markets that are local markets where farmers sell fresh produce directly to consumers. These markets support local economies and provide fresh, often organic products.

2.1.1 The Dynamics of the Market

- **Fragmented Supply Chains** - Small-scale farmers often deal with numerous intermediaries, including traders and middlemen, which can reduce their profit margins and complicate market access.
- **Limited Market Access** - Small-scale farmers typically sell their produce in local or regional markets, which may not always offer the best prices or access to broader consumer bases.
- **Price Volatility(Fluctuating Prices)**- Prices for agricultural products can be highly volatile due to seasonal variations, weather conditions, and market demand, affecting the income stability of small-scale farmers.
- **Infrastructure Challenges**- Poor Infrastructure, Inadequate transportation, storage, and market facilities often lead to inefficiencies and post-harvest losses, impacting the profitability of small-scale farming.
- **Credit and Financial Services** - Difficulty in accessing affordable credit and financial services can hinder farmers' ability to invest in inputs, technology, and infrastructure.
- **Regulatory and Policy Environment** - Navigating agricultural policies, subsidies, and market regulations can be challenging, particularly for small-scale farmers with limited resources.
- **Demand and Supply Imbalances(Market Demand)7.** - Small-scale farmers may struggle with demand forecasting and adjusting their production to match market needs, leading to surpluses or shortages.

2.2 Business Needs

Agriculture business refers to various commercial activities related to the cultivation of crops, raising of livestock, and processing of agricultural products. It encompasses a wide range of activities and opportunities across the agricultural value chain. Agriculture businesses are vital to global food systems, providing essential products and services while driving economic growth and sustainability. They offer diverse opportunities for entrepreneurs and established businesses to contribute to and benefit from the agricultural sector.

Market Demand: Understanding market trends and consumer preferences to align production with demand.

Market Access: Improved productivity and quality can help farmers meet market demands and gain better prices for their produce.

Technology Adoption: Leveraging technology for improved productivity, tools to optimize resource usage (water, fertilizers, pesticides) to reduce costs and environmental impact, for efficiency and sustainability.

Investment and Financing: Securing funding and managing financial resources for operations and growth.

Risk Management: Real-time data on weather patterns and pest outbreaks can help farmers mitigate risks and make timely decisions.

Yield Enhancement: Personalized advice based on specific farm conditions can lead to better crop management practices, increasing yields and profitability.

2.3 Customer Requirements

- **Precision Agriculture-** Recommendation of appropriate crops to grow based on soil nutrients, AI based technologies that analyze data from sensors, drones, and satellites to monitor soil moisture, crop health, and weather conditions in real time.
- **Crop and Soil Health Management-** AI based devices to monitor crop conditions and analyze soil data to recommend appropriate fertilization, irrigation, and soil management practices.
- **Irrigation and Water Management-** Optimize irrigation schedules based on real-time weather forecasts, soil moisture levels, and crop needs.
- **Early Pest Detection-** Identification of plant diseases or pests infestations and provide treatment recommendations.
- **Weather Forecasting and Alerts-** Accurate localized weather forecasts to help farmers plan their activities and protect crops from adverse conditions. Like in times of drought they can prepare for irrigation facilities beforehand or during upcoming rainfall they can reduce irrigation.
- **Market Insights and Pricing-** Knowledge of market trends and historical data to forecast future prices of crops and livestock.
- **Educating farmers on new techniques of farming** - Enlightenment of new techniques like cover cropping and reduced tillage to capture and store carbon in the soil, or cross-plantation techniques.
- **Personalized advices on agriculture and Multilingual Support-** A user-friendly interface that is easy to navigate and accessible to users with varying levels of technological expertise and 24/7 virtual assistance in multiple languages to cater to diverse user groups.

2.3.1 Customer Segmentation

- **Small-scale Farmers-** Typically have limited land and resources. They might need solutions for optimizing crop yield, managing resources, and accessing market information.
- **Medium-Scale Farmers-** They might be interested in more advanced techniques like precision farming, soil health monitoring, including predictive analytics, supply chain management and efficient resource allocation.
- **Organic Farmers** - Emphasize organic practices and may require specific solutions for soil health, natural pest control, and organic certification management.
- **Farm Management and Technology Providers** - Farm Management and Technology Providers would need the app to integrate and streamline data from various sources, offering comprehensive analytics and decision-making tools to enhance farm productivity and efficiency.
- **Agribusinesses** - Companies involved in the supply chain of agricultural products, from seed and fertilizer providers to food processors. They benefit from improved crop yields and quality.
- **Government** - The app can look forward to have a partnership with Government in order to make its reach to the farmers as much as possible. It can offer the Government with scalable solutions for monitoring and improving agricultural productivity across regions, providing real-time data and insights to support policy-making and resource allocation for sustainable farming practices.

- **Agricultural Research Firms** - The app can target agricultural research firms by providing a platform for real-time data collection and analysis, facilitating advanced research on crop performance, soil health, and pest management.

3.Target Specification

- Real-Time Monitoring:** FarmoAI will offer comprehensive monitoring of soil moisture, temperature, humidity, and crop health using IoT sensors. Farmers will receive real-time updates via mobile or web platforms, helping them adjust their actions as needed.
- Personalized Recommendations:** The app will provide crop-specific advice based on soil and weather data, including irrigation schedules, fertilizer recommendations, and pest control alerts.
- Pest and Disease Detection:** Using AI-driven image recognition, FarmoAI will identify pests and diseases from images taken by farmers, providing timely treatment suggestions to minimize damage.
- Weather Forecasting:** Localized weather forecasting will help farmers plan their activities, such as planting, irrigation, and harvesting, in response to upcoming weather conditions.
- Market Access:** The app will provide farmers with up-to-date market prices, helping them sell their produce at the best possible rates and maximize profits.
- Multilingual Support and Ease of Use:** Designed with a user-friendly interface, the app will include multilingual support to cater to a diverse group of users with varying technological expertise.
- Financial Tools:** Integration with financial institutions will allow farmers to apply for crop loans and insurance directly through the app, simplifying access to critical financial services.
- Sustainability:** The app will promote sustainable farming practices, offering recommendations that minimize resource waste and encourage environmentally friendly techniques.



4.External Search

4.1 Benchmarking Alternate Products

Existing Products

1. FarmLogs :

- **Overview:** FarmLogs is a comprehensive farm management app that provides tools for crop planning, field mapping, and financial tracking, helping farmers optimize their operations.
- **Strengths: Real-time Data Analytics:** FarmLogs excels in offering real-time data on weather conditions, soil health, and crop status, allowing farmers to make informed decisions.
- **Weaknesses: Limited Localization:** The app is primarily designed for the U.S. market, limiting its usefulness for farmers in other regions due to a lack of localized farming advice and language options.

2. AgriCentral

- **Overview:** AgriCentral is a smart farming app tailored to the needs of Indian farmers, offering features such as crop care, market prices, and weather forecasts.
- **Strengths: Market Integration:** AgriCentral provides real-time market prices and trends, helping farmers sell their produce at optimal rates, which is particularly beneficial in the Indian context.
- **Weaknesses: Complex Interface:** The app's extensive features and complicated interface can make it overwhelming for less tech-savvy users, and its interface could be more intuitive. It does not have image analysis feature for soil analysis or plant disease detection.

3. FarmBee

- **Overview:** FarmBee specializes in helping farmers grow high-quality pomegranates by offering personalized recommendations and farm management tools based on data analytics.
- **Strengths: Precision Agriculture:** The app provides highly accurate and data-driven insights specific to pomegranate cultivation, making it a valuable tool for farmers in this niche.
- **Weaknesses: Limited Crop Focus:** Its narrow focus on pomegranates limits its appeal to farmers growing other crops, restricting its potential user base and has no interface or chatbots for user interactions.

4. Agremo

- **Overview:** Agremo is a precision agriculture app that uses drone and satellite imagery to provide detailed crop analysis. It helps farmers monitor plant health, detect pests and diseases, and assess crop damage, offering actionable insights for better farm management.
- **Strengths:** Provides highly accurate and detailed aerial crop analysis, allowing for early detection of issues such as pests, diseases, and water stress.
- **Weaknesses:** Does not recommend fertilizers and limits its accessibility for smaller farms or those without advanced technology and is not affordable for small-scale farmers. It does not provide real-time alerts or notifications to the farmers regarding any issues.

5. Applicable Patents

Some applicable patents include:

5.1 Precision agriculture Management System

US Patent 10,394,415 B2 - “*Precision Agriculture Management System*” :

This patent covers a system for managing precision agriculture, including the use of sensors, data collection, and decision-making algorithms for crop management.

5.2 Precision Farming System

US Patent 9,594,260 B2 - “*Precision Farming System*”:

This patent relates to a system for precision farming that integrates GPS, soil sensors, and variable-rate technology to improve farming efficiency.

5.3 Smart control/IoT system for agriculture environment control

US20150342452A1 - “*Smart control/iot system for agriculture environment control*”

Systems and methods for smart farming using machine learning and data analytics.

5.4 Smart Irrigation System

US Patent 10,413,630 B2 - "*Smart Irrigation System Using Machine Learning*":

This patent involves a machine learning-based irrigation system that adjusts water usage based on historical data, weather conditions, and real-time sensor input.

US Patent 10,640,294 B2 - "*Artificial Intelligence-Driven Irrigation System*":

This patent describes an AI-driven system that optimizes water delivery by learning from various data sources such as weather patterns, soil conditions, and crop growth stages.

5.5 Pest-Detection System

US Patent 10,221,466 B2 - "*System and Method for Image-Based Pest Detection in Agriculture*" This patent describes a system that uses image processing techniques to detect pests on crops. The system can analyze images captured by cameras or drones and identify pests using machine learning algorithms.

6. Applicable Regulations

Some applicable regulations include:

6.1 Data Protection and Privacy Laws

Information Technology (Reasonable Security Practices and Procedures and Sensitive Personal Data or Information) Rules, 2011: Under the IT Act, 2000, these rules govern the collection, storage, and processing of personal data. If your app collects or processes data from users (e.g., farmers), you must comply with these rules.

6.2 Agricultural Laws and Standards

Fertilizer Control Order (FCO), 1985: If your app involves the recommendation of fertilizers, you must comply with this order, which regulates the quality, production, and distribution of fertilizers in India.

Essential Commodities Act, 1955: This act regulates the production, supply, and distribution of essential commodities, including agricultural produce. Ensure that your app does not promote practices that might violate this act, such as hoarding or illegal trading.

6.3 Intellectual Property Rights (IPR)

Patents Act, 1970: If your smart farming app includes innovative technology, consider filing for patents to protect your intellectual property. Similarly, be aware of existing patents to avoid infringement.

Trademarks Act, 1999: Protect your app's name, logo, and brand by registering them as trademarks under this act.

6.4 Contract Farming and Agribusiness Regulations

The Farmers (Empowerment and Protection) Agreement on Price Assurance and Farm Services Act, 2020: If your app facilitates contract farming or provides advisory services, ensure compliance with this act, which governs agreements between farmers and buyers.

6.5 e-Commerce Regulations (if applicable)

Consumer Protection Act, 2019: If your app includes a marketplace or sells products/services, ensure compliance with this act, which protects consumer rights and ensures fair trade practices.

7. Applicable Constraints

- **Budget Constraints:** Development and operational costs include software creation, hardware integration (IoT, sensors), cloud storage, and marketing efforts to promote the app.
- **Expertise Constraints:** A multidisciplinary team is required, including data scientists for predictive models, software developers for app creation, and agricultural experts for accurate recommendations.
- **Technology Constraints:** Reliable internet connectivity is essential, so offline functionality or alternative data methods are crucial in low-connectivity areas.
- **Scalability:** The app must be designed to scale with growing user numbers and data volumes, requiring scalable cloud infrastructure.
- **Logistic Constraints:** Effective distribution in remote areas is necessary, alongside plans for the sourcing, installation, and maintenance of hardware devices like IoT sensors.

8. Feasibility: Short-Term Development (2-3 years)

Farmologist is highly feasible to develop within the next 2-3 years due to rapid advancements in agricultural technology, AI, and IoT devices. The infrastructure required for AI-based farming solutions, such as cloud computing and smart sensors, is already widely available and becoming more affordable. Initial product development will focus on:

- **Data collection** from various farms through IoT sensors for soil moisture, weather, and crop conditions.
- **AI algorithms** for predictive farming analytics, yield optimization, and crop health monitoring.
- **Mobile or web platforms** to deliver actionable insights to farmers in real-time.

Given the increasing adoption of digital tools in agriculture, especially by large-scale farms and cooperatives, FarmoAI can be developed within the next two years with a phased rollout of features.

9. Viability: Long-Term Relevance (20-30 years)

Farmologist is positioned to be highly viable for the long term due to the growing global focus on sustainable agriculture, food security, and climate change resilience. With increasing population demands and limited arable land, farmers will need AI-driven solutions to maximize crop yield and efficiency. Over the next 20-30 years:

- AI will play a central role in precision farming, addressing global challenges such as water shortages, changing weather patterns, and soil degradation.
- As climate change intensifies, real-time data and predictive models offered by FarmoAI will be crucial for adaptive farming techniques.
- The platform can evolve to incorporate emerging technologies like drone-based data collection, genetic crop optimization, and advanced robotics, ensuring it remains relevant and ahead of competitors.

10. Monetization Strategies / Business Model

Freemium Model: Provide a free version of the app with basic features and charge for advanced features or enhancements. This model can help attract users and convert them into paying customers over time.

Weather forecasting feature, virtual assistance/ Chatbot can be provided for free.

The tutorials or training videos of the app and other videos for farming knowledge

Basic information on best time to plant and harvest, importance of crop monitoring and detecting of pest should be provided to farmers for free.

Subscription Model: We can charge fees based on annual (yearly) or monthly. There can be three tiers.

Basic Tier: Offer essential features at a low cost. This might include basic data monitoring and simple analytics like pest detection, soil analysis etc.

Standard Tier: Include more advanced features like detailed analytics, custom reports, and additional data sets including crop recommendation, irrigation schedules, fertilizers recommendation.

Premium Tier: Provide top-tier features such as real-time alerts, advanced predictive analytics, expert consultations, financial insights of farmers, yield and market prediction etc.

Free Trial : Offer a free trial period for new users. This allows them to experience the premium features before committing to a subscription.

In-App Purchases: Provide additional features, tools, or data sets that users can purchase within the app. This could include detailed reports, expert consultations, or additional crop data.

Farmers can access specific features or detailed reports on a pay-per-use basis.

This is useful for farmers who need occasional access to any expert advice or other features.

Advertising :

In-App Ads: Display ads within the app ensuring they don't disrupt the user experience. Targeted ads based on user activity can be effective.

Sponsored Content: Partner with agricultural companies to feature their products or services in the app.

Partnerships and Sponsorships: Collaborate with agricultural companies, seed manufacturers, or equipment suppliers to offer sponsored content or features within the app.

Agricultural-Tech Companies

Integration Partnerships: Partner with companies that provide agricultural technology (e.g., IoT sensors, drones). Integrate their hardware or services with your app and share revenue.

Co-Branding: Collaborate on co-branded marketing campaigns or joint product offerings. This could include special offers or bundled solutions.

Government Agencies:

Public Programs: Partnership with government agencies on agricultural programs or initiatives. They might provide funding or support in exchange for access to your app's technology and data.

Government help or assistance is very important as it can help in making this app popular on national level and improve technology in rural areas so as to make this app more feasible.

Agricultural Suppliers:

Product Recommendations: Partner with suppliers of seeds, fertilizers, or equipment. Feature their products in your app and earn a commission on sales generated through your app.

Sponsored Content: Include sponsored content or articles about their products in your app's educational section.

Research Institutions

Data Collaboration: Collaborate with research institutions for data sharing . Your app's data can provide valuable insights, while they can offer research support, funding, advanced information on farming beneficial for the app.

7 . Affiliate Marketing

Agricultural Products: Including agricultural services or products and promoting them through the app and earn commissions on sales.

Technology Solutions: Partner with tech companies that offer complementary solutions, such as data analytics tools or farming equipment.

Product Listings: Create a marketplace or product recommendations section in your app where users can browse and purchase affiliated products.

8 . Training and Workshops: Workshops, educational content, online courses on the app can be given by providing hands-on training sessions for farmers, agribusinesses, or agricultural students. These workshops could focus on teaching participants how to use the app effectively, understand data analytics, implement smart farming techniques, and integrate with IoT devices. Charge a fee for participation, offering different pricing tiers for basic, advanced, or specialized sessions. Additionally, we could offer certificates of completion to add value and attract more attendees.

9 . Partnering with Financial Institutions : Collaborate with banks or microfinance institutions to offer financial products like crop insurance or loans, earning a commission for each referral or transaction processed through the app.

10 . Data Licensing : Data collected from the platform can be anonymized and licensed to research institutions, agricultural companies, and government bodies for analysis of trends and insights.

9.Concept Generation

I was inspired to create a technological solution to assist farmers by leveraging AI and Machine learning because I believe technology should play a pivotal role in improving the lives of rural communities. Farmers are the backbone of our nation, working tirelessly to ensure food security and contributing significantly to the economy. However, despite their hard work, many farmers struggle with low income and limited access to resources. By developing tools that harness the power of AI and ML, we can provide farmers with the insights and support they need to increase productivity, optimize resources, and ultimately improve their livelihoods. This approach not only empowers farmers but also promotes sustainable agricultural practices, ensuring that technology serves the people who need it most.

The idea of creating a technological solution for farmers came to me out of a deep respect for the hardworking individuals who sustain our nation's food supply, yet often struggle to make ends meet. The vision is to use technology as a bridge to uplift rural communities, ensuring that those who feed us can also thrive economically.

10. Concept Development

App Name: Farmologist

“Farmologist” is an innovative mobile application designed to revolutionize agricultural practices by leveraging artificial intelligence and machine learning. The prototype represents a functional version of the app, focused on providing farmers with essential tools and insights to optimize their operations, enhance productivity, and promote sustainability.

An insight on a possible interaction of a farmer and the app --

Farmer Logs In: A farmer logs into the app in the morning and checks the dashboard, which shows a weather alert for potential frost that night.

App Suggests Action: The app recommends covering sensitive crops and adjusting irrigation schedules to minimize frost damage. A reminder is also set for alerting him to do the same.

Market Insights: The farmer also checks market insights and sees that demand for a particular crop is expected to rise next week. The app suggests waiting a few days before selling to get a better price.

Crop Health Check: The farmer notices some yellowing leaves on a crop. They take a picture, upload it to the app, and the app diagnoses a nutrient deficiency, suggesting a specific fertilizer.

Community Support: Later, the farmer posts in the community forum, asking for advice on an unusual pest they spotted. Another farmer responds with a solution they've used successfully.

11. Financial Equation

Assumptions:

1. Revenue Sources:

- Freemium model with premium subscription features.
- In-app purchases (consultations, advanced features).
- Partnerships (telemedicine, insurance, and agri-pharma).
- Targeted advertising.

2. Cost Inclusions:

- Development and infrastructure (personnel, servers, maintenance).

- Marketing and customer acquisition.
- Legal and compliance costs.
- Operational expenses (cloud services, data processing).

Financial Equation Framework:

$$\text{Profit (P)} = \text{Total Revenue (R)} - \text{Total Costs (C)}$$

Where:

- **Total Revenue (R)** = Revenue from subscriptions + In-app purchases + Partnerships + Ads
- **Total Costs (C)** = Development costs + Marketing + Legal/compliance + Operational costs

Revenue (R):

1. Freemium Subscription (R_s):

- Subscription cost per user per month = S_p
- Number of paying users = N_p
- Number of free users = N_x
- Premium conversion rate = P_a
- **Annual Subscription Revenue:** $R_s = N_x \times P_a \times S_p \times 12$

2. In-App Purchases (R_{iap}):

- Average purchase per user = P_{iap}
- **Annual In-App Purchase Revenue:** $R_{iap} = N_p \times P_{iap} \times 12$

3. Partnership Revenue (R_p):

- K_p is a constant representing the percentage of total users engaged in partnerships.
- **Partnership Revenue:** $R_p = N_x \times K_p$

4. Advertising Revenue (R_a):

- Revenue per ad impression = A_i
- Average impressions per user = I_i
- **Advertising Revenue:** $R_a = N_x \times I_i \times A_i$

Costs (C):

1. Development and Infrastructure (C_x):

- Annual cost for development, hosting, and maintenance = C_x

2. Marketing (C_m):

- Cost per user acquisition = C_{ma}
- Growth rate = G

3. Marketing Cost: $C_m = C_{ma} \times N_x \times G$

4. Legal and Compliance (C_l):

- Annual legal/compliance costs = C_l

Final Profit Equation:

$$P = (N_x \times P_a \times S_p \times 12) + (N_p \times P_{iap} \times 12) + (N_x \times K_p) + (N_x \times I_i \times A_i) - (C_x + C_{ma} \times N_x \times G + C_l)$$

Growth Rate Equation:

Let:

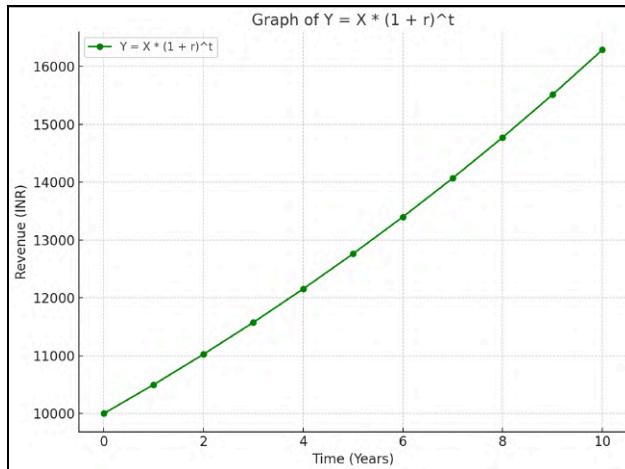
- Y = Total profit over time
- X = Initial revenue
- r = Growth rate (5%)
- t = Time in years

The equation for growth over time is: $Y=X\times(1+r)^t$

Example:

- If initial revenue is INR 10,000 and the growth rate is 5% over 3 years:

$$Y=10,000\times(1.05)^3=10,000\times1.157625=\text{INR } 11,576.25$$



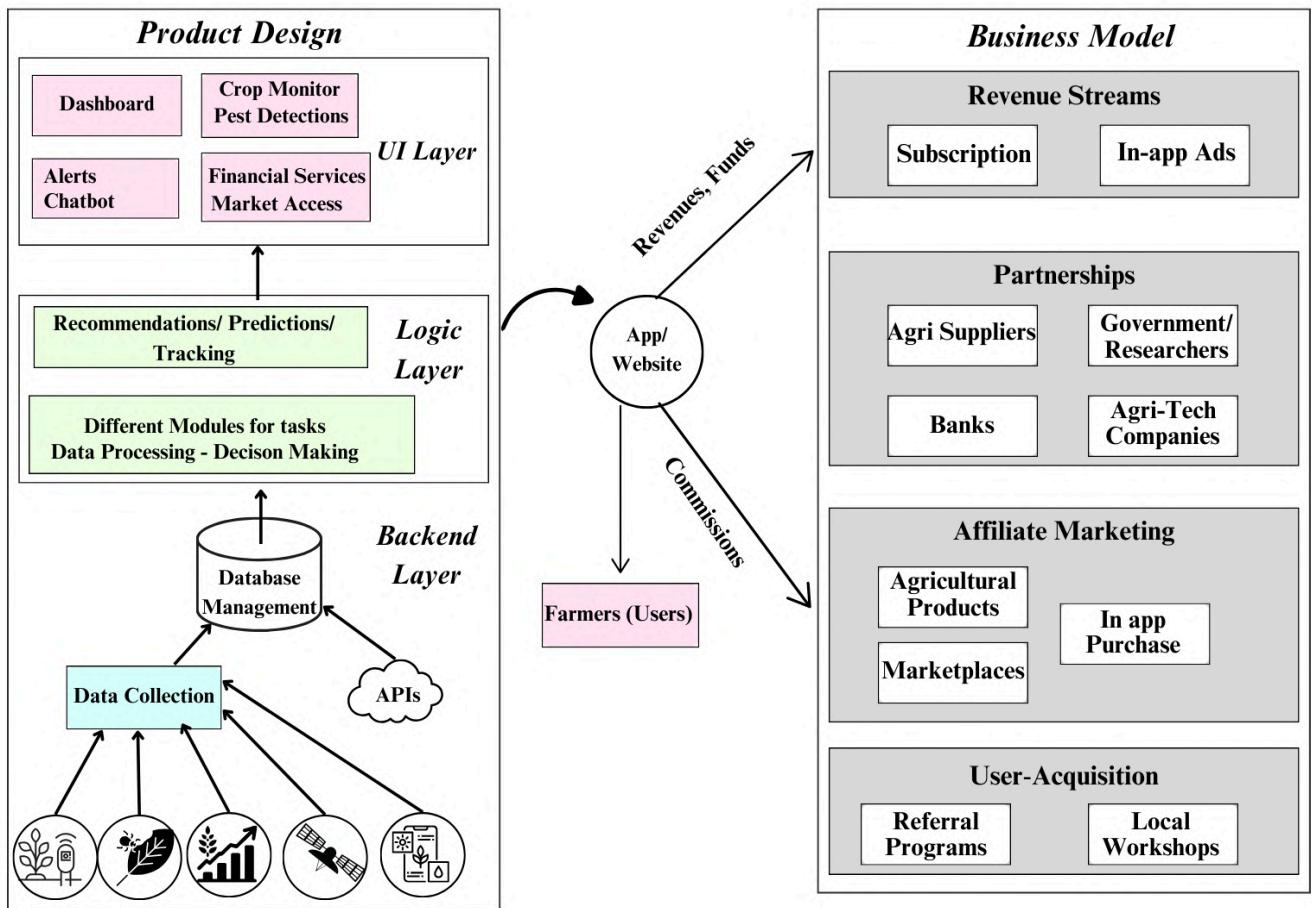
11.Final Product Prototype

The Farmologist prototype is a user-friendly app that offers key features such as crop health monitoring, localized weather forecasts, financial services integration for loan applications, and real-time market prices. The interface is designed for easy navigation, allowing farmers to access personalized farming tips and tools with minimal effort. Initial testing focuses on usability, ensuring the app meets the needs of rural farmers effectively.

- **Product Prototype:** Farmologist begins with **data collection** through IoT sensors, manual input, and third-party APIs (weather, market prices). This data is **processed** using modules for data cleaning, integration, and analysis, with machine learning providing predictive insights and personalized recommendations. The processed data is stored securely in databases, with **real-time processing** enabling immediate alerts and actions. The **user interface** is designed to be intuitive, featuring dashboards for crop management, financial services, and market access, all accessible via a mobile app. This interface allows farmers to easily navigate and utilize the app's features, receiving actionable insights and making informed decisions.
- **Business Model:** Farmologist's revenue comes from **subscription fees** for advanced features, **transaction fees** from crop sales through the app, and **data monetization** from anonymized data insights. Partnerships with **financial institutions** and **Agri-input suppliers** enhance the app's value, offering loans and resources directly through the platform. The app will launch in targeted rural regions, using **local partnerships** and **government collaborations** for effective user acquisition, ensuring widespread

adoption among small to medium-scale farmers.

Schematic diagram:



Schematic diagram of the final product prototype

12. Product Details

12.1 How does it work?

1. User Onboarding and Setup

- App download and register:* Farmers download the app and register using their mobile number, email, or social media accounts.
- Set up profile :* They set up a profile, providing details about their farm, including location, crop types, farm size, and resources available (e.g., water supply, equipment).
- Farm Mapping and Configuration:* The app may request farmers to map their farm fields using GPS or manual input. Users can configure settings such as preferred language, notification preferences etc.

2. Regular Interaction

- Daily Usage:* Farmers use the app daily to check weather forecasts, monitor crop health, and receive recommendations.
- Notifications and Alerts:* The app sends push notifications or SMS alerts for critical events, such as upcoming adverse weather, pest infestations, or market price changes.
- Manual Data Entry :* Farmers can manually input data such as planting dates, crop observations, and resource usage (water, fertilizers).

- *Community Forum*: Farmers interact with other users and agricultural experts through the app's community forum. They can ask questions, share experiences, and receive advice, creating a collaborative learning environment.
- *Expert Consultations*: The app may offer access to agricultural experts who can provide personalized advice through chat, video calls, or scheduled consultations.

3. Interaction with Market Platforms

- The app integrates with market data platforms to pull real-time information on crop prices.
- This data is used to provide farmers with insights on the best times to sell their produce.
- The app may analyze market trends to forecast demand for certain crops, helping farmers make informed decisions on crop selection and timing.
- Marketplace Integration: The app may connect farmers directly with online marketplaces or local buyers.

4. Interaction with Agricultural Experts and Organizations

- The app may offer live chat or video consultations with agricultural experts.
- Farmers can schedule appointments or get instant advice on pressing issues.
- Agricultural experts may contribute articles, videos, and tutorials to the app's knowledge base.
- Collaboration with Agricultural Institutions and Research & Development.

5. Bank Interaction for Crop Finance Loans:

- The app facilitates access to crop finance loans by connecting farmers with partnered banks, allowing them to apply directly through the app with pre-filled farm data and digital submission of required documents.

12.2 Data Sources

1. **Weather Data Providers** : The app integrates with weather APIs to receive real-time and forecasted weather data specific to the farm's location.
2. **Government and NGO Databases** : The app may connect with government or NGO databases to access agricultural research, best practices, and local regulations.
3. **Sensor Data Collection**: If the farm is equipped with IoT devices (e.g., soil moisture sensors, weather stations), the app collects data from these devices in real-time.
4. **Manual- Data Collection** : Farmers can manually input data such as planting dates, crop observations, and resource usage (water, fertilizers). Farmers can upload images of crops for AI-based pest and disease detection.
5. **Satellite Imagery** : for crop health monitoring and land use analysis.
6. **Market Data** : Market Price APIs providing real-time crop prices, demand trends, and forecasts.
7. **Financial Data** : Bank APIs for loan processing, interest rates, and financial support tailored to agricultural needs.
8. **Agricultural Institutions and Universities** for the latest research, training content, and expert advice.

12.3 Algorithms, Frameworks

12.3.1 Algorithms

- **Regression Models** for yield prediction and market price forecasting.
- **Classification Models** (e.g., Decision Trees, Random Forest) for pest and disease detection based on image analysis and for recommendation systems like crop and fertilizers recommendation.

- **Clustering Algorithms** (e.g., K-Means) for segmenting farms based on similar characteristics (soil type, climate).
- **NLP** : For virtual assistance or Chatbot.
- **Convolutional Neural Networks (CNNs)** for analyzing images of crops to detect diseases, pests, and nutrient deficiencies.
- **Linear Programming** for optimizing resource usage (water, fertilizers).
- **Genetic Algorithms** for crop planning and scheduling to maximize yield.

12.3.2 Frameworks

- **Deep Learning Frameworks:** **TensorFlow** or **PyTorch** for building and training AI models for image recognition and predictive analytics.
- **Web and Mobile Application Frameworks:** **React Native** or **Flutter** for cross-platform mobile app development.
 - **Django** or **Node.js** for backend development and API management.
- **Data Processing and Analytics:**
 - **Apache Spark** for large-scale data processing and real-time analytics.
 - **Pandas** and **NumPy** for data manipulation and analysis.
- **Data Visualization:**
 - **Tableau** or **Power BI** for creating dashboards and visualizing farm data.
 - **Matplotlib** and **Seaborn** for custom data visualizations within the app.

12.4 Teams Required

12.4.1 Product Development

Full-Stack Developer: Handles both frontend and backend development.

Data Scientist/AI Specialist: Develops and implements machine learning algorithms for crop prediction, pest detection, and market analysis.

UI/UX Designer: Designs user-friendly interfaces and ensures a smooth user experience.

12.4.2 Operational Development

- **IoT Specialist :**
Integrates IoT devices with the app for real-time data collection and monitoring.
Manages sensor data and its visualization.
- **DevOps Engineer** : Sets up cloud infrastructure, CI/CD pipelines, and ensures the app's scalability.

12.4.3 Support Team

- **Agricultural Expert (Consultant):**
 - Provides domain expertise, ensuring that the app's recommendations are accurate and relevant.
 - Can be brought in on a consulting basis.
- **Customer Support (Part-time):**
 - Handles user queries, offers guidance, and collects feedback.
 - Important for maintaining user satisfaction and retention.
- **Quality Assurance Tester(Part-time):**
- Ensures the app is bug-free and functions well across different devices.

12.4.4 Business Development

- Partnership Managers: To establish and maintain relationships with managers of the app's online presence, user acquisition strategies, and partnerships with banks, markets, and agricultural institutions.
- Focuses on cost-effective digital marketing strategies.
- Marketing Specialists: To attract new users and promote the service.

12.5 Costs

1. Initial Development Costs:

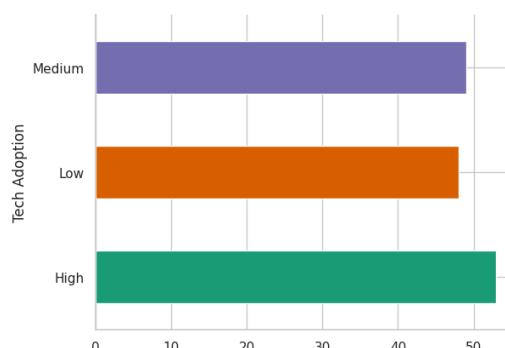
- App Development: For building and launching the mobile app.
- Backend Infrastructure: For setting up servers, databases, and APIs.
- Design: For UX/UI design.
- AI/ML Integration: Developing algorithms for crop predictions, recommendations, etc.
- IoT Integration: Connecting sensors and devices for real-time monitoring.
- Initial Support: Addressing user feedback and making post-launch updates

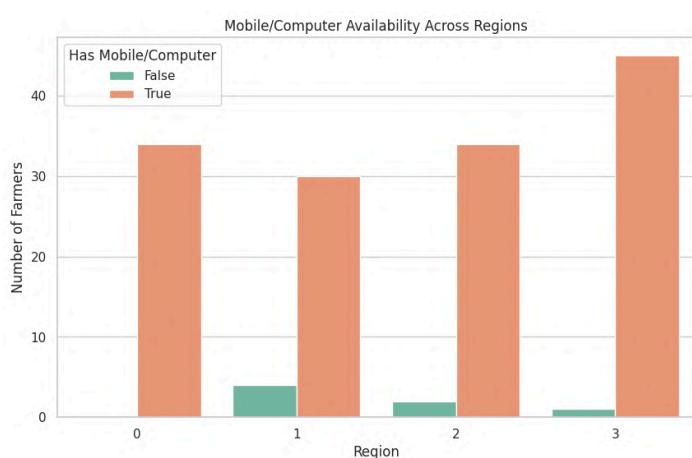
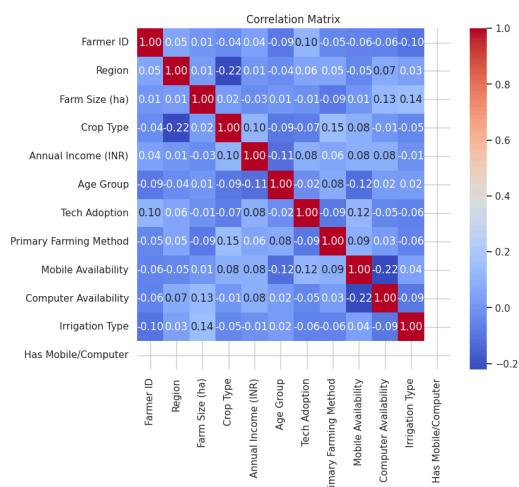
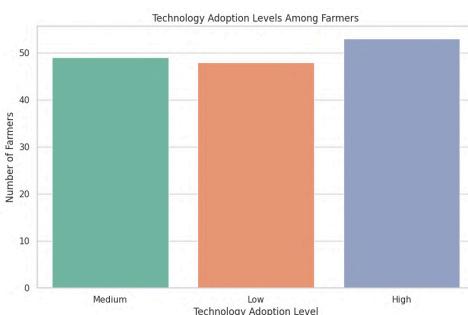
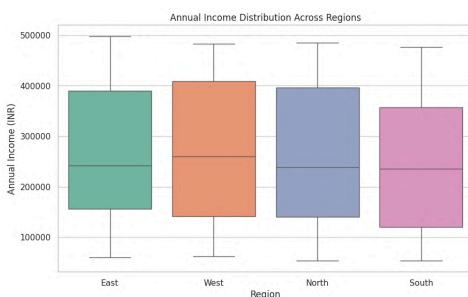
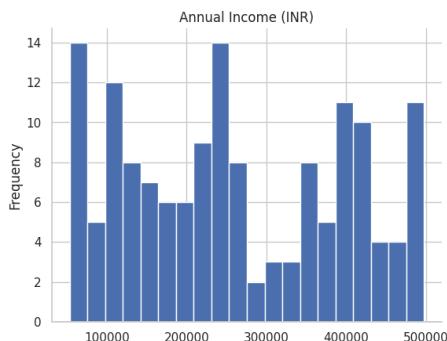
2. Ongoing Operational Costs:

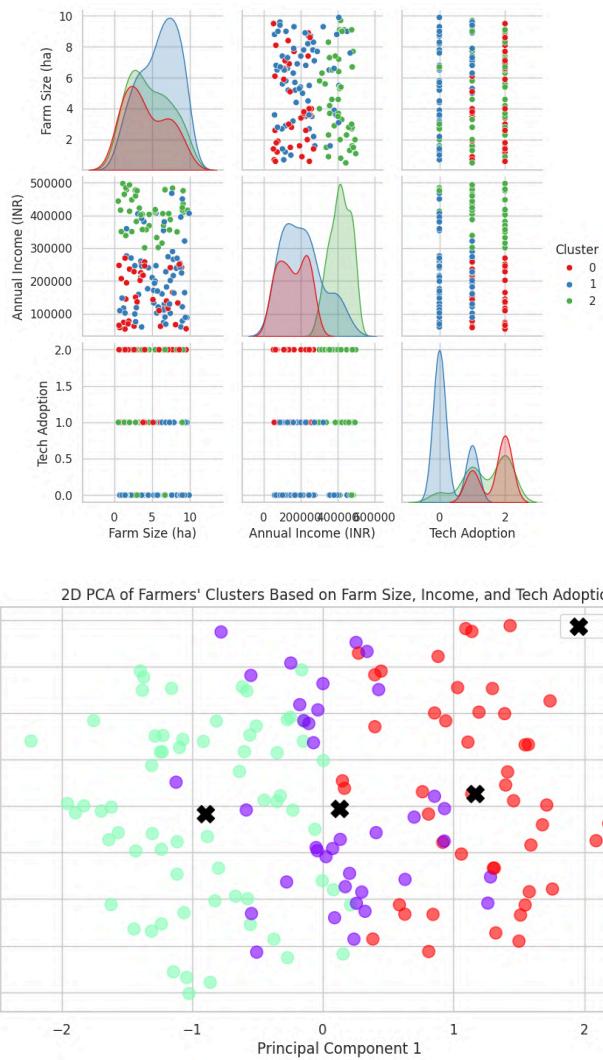
- Salaries: For a full team including developers, quality checkers, agricultural experts.
- App Store Submission: Preparing and submitting the app to Google Play and Apple App Store paying fee and for continuous updates.
- Technology Maintenance: For server costs, software updates, and security.
- Marketing Campaigns: Digital marketing, social media promotion, and outreach.
- Tutorials : Creating tutorials, guides, and support resources.

13. Market Segmentation : [github](#)

Market segmentation analysis for a farming app helps identify distinct groups of farmers based on characteristics such as farm size, annual income, and technology adoption. By applying clustering techniques like K-Means, farmers can be categorized into segments that share similar attributes, which allows the app to deliver more personalized solutions. For example, insights from segmentation can inform targeted advisory services, tailored financial products, or region-specific agricultural techniques. This segmentation improves engagement and satisfaction by ensuring that farmers receive information and resources most relevant to their specific needs.







14. Small-Scale Code Implementation

Since developing the fully functional application from scratch is cumbersome as it involves numerous steps, data collection, deployment etc. To make a simpler process I've prepared a small-scale implementation of a **Crop Recommendation System** that has been developed using soil nutrients and weather as inputs. The other functionalities like pest detection, irrigation alerts, chatbot etc. are excluded. The dataset used is a pre-defined dataset by Atharva Ingle from Kaggle.

Dataset : [Crop Recommendation Dataset](https://www.kaggle.com/datasets/atharyaingle/crop-recommendation-dataset) (<https://www.kaggle.com/datasets/atharyaingle/crop-recommendation-dataset>)

1 . Data Preprocessing

Reading the csv file

```

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')

df=pd.read_csv('Crop_recommendation.csv')

df.head()

```

	N	P	K	temperature	humidity	ph	rainfall	label
0	90	42	43	20.879744	82.002744	6.502985	202.935536	rice
1	85	58	41	21.770462	80.319644	7.038096	226.655537	rice
2	60	55	44	23.004459	82.320763	7.840207	263.964248	rice
3	74	35	40	26.491096	80.158363	6.980401	242.864034	rice
4	78	42	42	20.130175	81.604873	7.628473	262.717340	rice

Missing Values and unique columns

```

df.isnull().sum()

df.duplicated().sum()

```

```

[10] df['label'].unique()

```

label	count
rice	100
maize	100
chickpea	100
kidneybeans	100
pigeonpeas	100
mothbeans	100
mungbean	100
blackgram	100
lentil	100
pomegranate	100
banana	100
mango	100
grapes	100
watermelon	100
muskmelon	100
apple	100
orange	100
papaya	100
coconut	100
cotton	100
jute	100
coffee	100

```

x=df.drop('label',axis=1)
y=df['label']

```

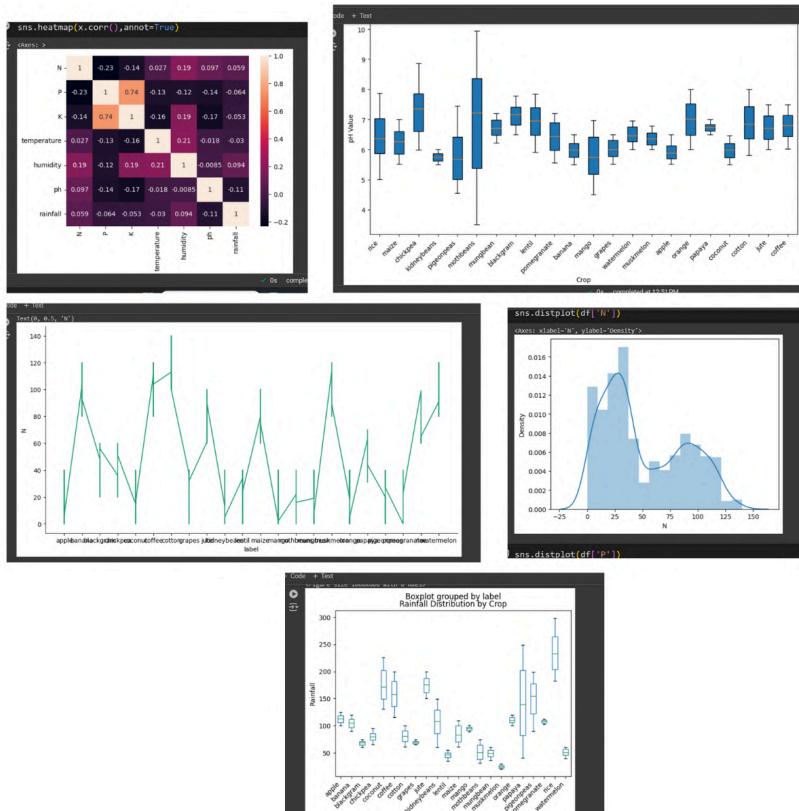
```

df['label'].value_counts()

```

label	count
rice	100
maize	100
chickpea	100
kidneybeans	100
pigeonpeas	100
mothbeans	100
mungbean	100
blackgram	100
lentil	100
pomegranate	100
banana	100
mango	100
grapes	100
watermelon	100
muskmelon	100
apple	100
orange	100
papaya	100
coconut	100
cotton	100
jute	100
coffee	100

2 . Exploratory Data Analysis



4 . Splitting dataset

```
Splitting the dataset
from sklearn.model_selection import train_test_split
Xtrain, Xtest, Ytrain, Ytest = train_test_split(x,y,test_size = 0.2,random_state =2)
[25] Xtrain
```

	N	P	K	temperature	humidity	ph	rainfall
1936	113	38	25	22.000851	79.472710	7.388266	90.422242
610	28	35	22	29.530376	86.733460	7.156563	59.872321
372	11	61	21	18.623288	23.024103	5.532101	135.337803
1559	29	139	205	23.641424	93.744615	6.155939	116.691218
1500	24	128	196	22.750888	90.694892	5.521467	110.431786

3 . Building the model

```
from sklearn.metrics import accuracy_score,classification_report,confusion_matrix
[29] from sklearn.tree import DecisionTreeClassifier
DecisionTree = DecisionTreeClassifier(criterion="entropy",random_state=2,max_depth=5)
DecisionTree.fit(Xtrain,Ytrain)

predicted_values = DecisionTree.predict(Xtest)
x = accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('Decision Tree')
print("DecisionTrees's Accuracy is: ", x*100)

print(classification_report(Ytest,predicted_values))
DecisionTrees's Accuracy is: 99.0
precision    recall  f1-score   support
```

```
Code Text Disk Geronimo
```

Logistic Regression

```
[x] [30] from sklearn.preprocessing import StandardScaler
  sc = StandardScaler()
  Xtrain_scaled = sc.fit_transform(Xtrain)
  Xtest_scaled = sc.transform(Xtest)

[x] [31]
  from sklearn.linear_model import LogisticRegression

  lr = LogisticRegression(multi_class = 'multinomial', solver='lbfgs')

  lr.fit(Xtrain_scaled,Ytrain)

  predicted_values = lr.predict(Xtest_scaled)

  x = accuracy_score(Ytest, predicted_values)
  acc.append(x)
  model.append('Logistic Regression')
  print("Logistic Regression's Accuracy is: ", x)
```

```
Code Text Disk Geronimo
```

Random Forest

```
[x] [32] from sklearn.ensemble import RandomForestClassifier
  rf = RandomForestClassifier(n_estimators=100, random_state=42)
  rf.fit(Xtrain, Ytrain)
  predicted_values = rf.predict(Xtest)
  x = accuracy_score(Ytest, predicted_values)
  acc.append(x)
  model.append('Random Forest')
  print("Random Forest's Accuracy is: ", x)
```

Random Forest's Accuracy is: 0.9954545454545455

```
Code Text Disk Geronimo
```

Naive Bayes

```
[x] [34] from sklearn.naive_bayes import GaussianNB

  NB=GaussianNB()

  NB.fit(Xtrain,Ytrain)
  predicted_values = NB.predict(Xtest)
  x =accuracy_score(Ytest, predicted_values)
  acc.append(x)
  model.append('Naive Bayes')
  print("Naive Bayes's Accuracy is: ", x)

  print(classification_report(Ytest,predicted_values))
```

Naive Bayes's Accuracy is: 0.9909090909090909
precision recall f1-score support

```
Code Text Disk Geronimo
```

Support Vector Machine

```
[x] [33] from sklearn.svm import SVC
  SVM = SVC(kernel='rbf', degree=3, C=1)
  SVM.fit(Xtrain_scaled,Ytrain)
  predicted_values = SVM.predict(Xtest_scaled)
  x =accuracy_score(Ytest, predicted_values)
  acc.append(x)
  model.append('SVM')
  print("SVM's Accuracy is: ", x)

  print(classification_report(Ytest,predicted_values))
```

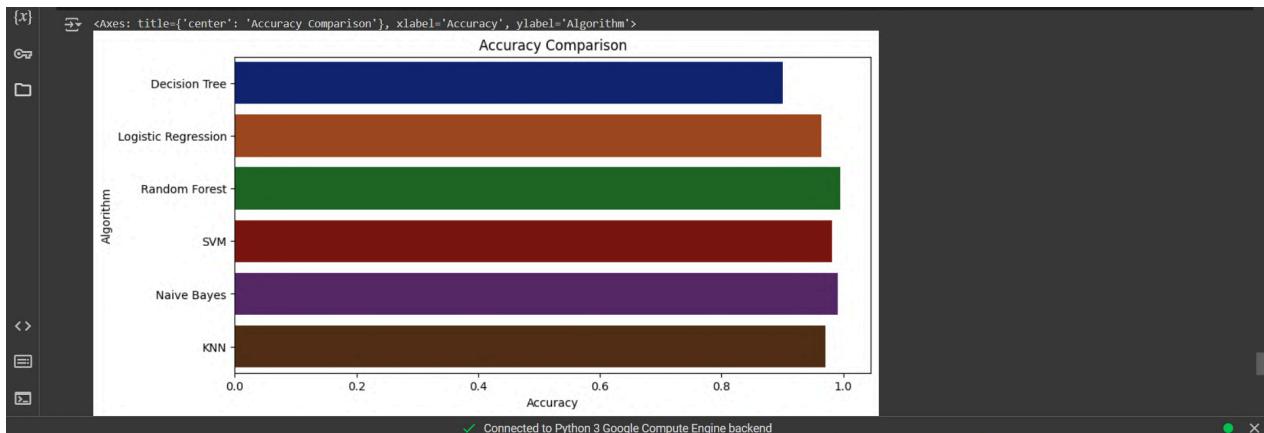
SVM's Accuracy is: 0.9818181818181818
precision recall f1-score support

```
Code Text Disk Geronimo
```

K-Nearest Neighbors

```
[x] [34]
  from sklearn.neighbors import KNeighborsClassifier
  knn=KNeighborsClassifier(n_neighbors=5)
  knn.fit(Xtrain_scaled,Ytrain)
  predicted_values = knn.predict(Xtest_scaled)
  x =accuracy_score(Ytest, predicted_values)
  acc.append(x)
  model.append('KNN')
  print("KNN's Accuracy is: ", x)
  print(classification_report(Ytest,predicted_values))
  conf = confusion_matrix(Ytest, predicted_values)
  print(conf)
```

KNN's Accuracy is: 0.9704545454545455
precision recall f1-score support



```
print(calculate(data, headers=[ 'Model' , 'Accuracy' ], tablefmt='pretty '))

[38] data = np.array([[90,42,43,20.87974371,82.00274423,6.50298529200001,202.9355362]])
prediction =rf.predict(data)
print("The crop most suitable to be cultivated is:")
print(prediction)

The crop most suitable to be cultivated is:
['rice']
```

• Using JobLib

```
import joblib

[42] filename='finalized_model'
joblib.dump(rf,'finalized_model')
['finalized_model']

[43] app=joblib.load('finalized_model')

[44] arr=[[104,18,30,23,60.3,6.7,140.9]]
y_pred=app.predict(arr)

[45] y_pred
array(['coffee'], dtype=object)
```

• Backend Using Flask

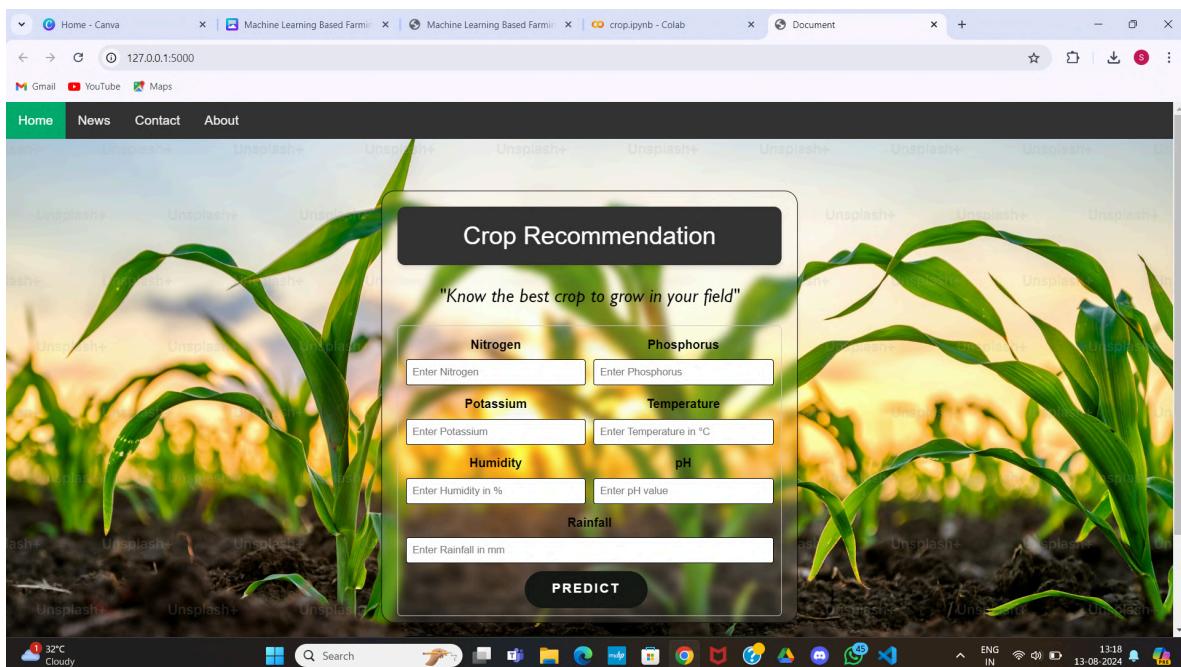
```
1 import joblib
2 from flask import Flask, render_template,request,redirect
3 # creating flask app
4 app = Flask(__name__)
5
6 @app.route('/')
7 def index():
8     return render_template("index.html")
9
10 @app.route("/predict",methods=['POST'])
11 def predict():
12     N = float(request.form['Nitrogen'])
13     P = float(request.form['Phosphorus'])
14     K = float(request.form['Potassium'])
15     temp = float(request.form['Temperature'])
16     humidity = float(request.form['Humidity'])
17     pH = float(request.form['pH'])
18     rainfall = float(request.form['Rainfall'])
19
20     feature_list = [N, P, K, temp, humidity, pH, rainfall]
21
```

```

11 def predict():
12     temp = float(request.form['Temperature'])
13     humidity = float(request.form['Humidity'])
14     pH = float(request.form['pH'])
15     rainfall = float(request.form['Rainfall'])
16
17     feature_list = [N, P, K, temp, humidity, pH, rainfall]
18
19     if pH>8 and pH<=14 and temp<100 and humidity>0:
20         joblib.load('finalized_model','r')
21         model=joblib.load(open('finalized_model','rb'))
22         arr=[feature_list]
23         res= model.predict(arr)
24         return render_template('index.html',prediction=res[0])
25
26     else:
27         return "Sorry, we could not determine the best crop to be cultivated with given parameters"
28
29
30
31
32

```

4 . API page



b. Crop Disease Detection

I have implemented a small-scale code for the model development of crop disease detection using Convolutional Neural Networks (CNN). The model was trained on the [PlantVillage](#) dataset, sourced from Kaggle, which contains images of various crops affected by different diseases. This small-scale

implementation focuses solely on the model-building phase, with CNN architecture tailored to detect patterns in the images for precise classification of diseased crops.

Dataset : <https://www.kaggle.com/datasets/abdallahhalidev/plantvillage-dataset>

1.Importing the Libraries

```
Crop_disease.ipynb
File Edit View Insert Runtime Tools Help All changes saved
Comment Share Gemini S
+ Code + Text
[ ] import os
import json
from zipfile import ZipFile
from PIL import Image
import matplotlib.image as mpimg

[ ] import tensorflow as tf
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

Double-click (or enter) to edit

● !pip install kaggle

Requirement already satisfied: kaggle in /usr/local/lib/python3.10/dist-packages (1.6.17)
Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.10/dist-packages (from kaggle) (1.16.0)
Requirement already satisfied: certifi>=2023.7.22 in /usr/local/lib/python3.10/dist-packages (from kaggle) (2024.8.30)
Requirement already satisfied: python-dateutil in /usr/local/lib/python3.10/dist-packages (from kaggle) (2.8.2)
Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from kaggle) (2.32.3)
Requirement already satisfied: tqdm in /usr/local/lib/python3.10/dist-packages (from kaggle) (4.66.5)
Requirement already satisfied: python-slugify in /usr/local/lib/python3.10/dist-packages (from kaggle) (8.0.4)
Requirement already satisfied: urllib3 in /usr/local/lib/python3.10/dist-packages (from kaggle) (2.0.7)
```

2>Loading the dataset from Kaggle

```
+ Code + text
[ ] !pip install kaggle

Requirement already satisfied: kaggle in /usr/local/lib/python3.10/dist-packages (1.6.17)
Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.10/dist-packages (from kaggle) (1.16.0)
Requirement already satisfied: certifi>=2023.7.22 in /usr/local/lib/python3.10/dist-packages (from kaggle) (2024.8.30)
Requirement already satisfied: python-dateutil in /usr/local/lib/python3.10/dist-packages (from kaggle) (2.8.2)
Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from kaggle) (2.32.3)
Requirement already satisfied: tqdm in /usr/local/lib/python3.10/dist-packages (from kaggle) (4.66.5)
Requirement already satisfied: python-slugify in /usr/local/lib/python3.10/dist-packages (from kaggle) (8.0.4)
Requirement already satisfied: urllib3 in /usr/local/lib/python3.10/dist-packages (from kaggle) (2.0.7)
Requirement already satisfied: bleach in /usr/local/lib/python3.10/dist-packages (from kaggle) (6.1.0)
Requirement already satisfied: webencodings in /usr/local/lib/python3.10/dist-packages (from bleach->kaggle) (0.5.1)
Requirement already satisfied: text-unidecode>1.3 in /usr/local/lib/python3.10/dist-packages (from python-slugify->kaggle) (1.3)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests->kaggle) (3.3.2)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests->kaggle) (3.8)

● !kaggle datasets download -d abdallahhalidev/plantvillage-dataset

Dataset URL: https://www.kaggle.com/datasets/abdallahhalidev/plantvillage-dataset
License(s): CC-BY-NC-SA-4.0
Downloading plantvillage-dataset.zip to /content
100% 2.04G/2.04G [00:52<00:00, 40.5MB/s]
100% 2.04G/2.04G [00:52<00:00, 41.9MB/s]

[ ] with ZipFile("plantvillage-dataset.zip", 'r') as zip_ref:
    zip_ref.extractall()
```

The screenshot shows two Jupyter Notebook cells. The first cell contains Python code for exploring a dataset:

```
[ ] with ZipFile("plantvillage-dataset.zip", 'r') as zip_ref:  
    zip_ref.extractall()  
  
{x} [ ] print(os.listdir("plantvillage dataset"))  
  
[ ]  
print(len(os.listdir("plantvillage dataset/segmented")))  
print(os.listdir("plantvillage dataset/segmented"))  
  
print(len(os.listdir("plantvillage dataset/color")))  
print(os.listdir("plantvillage dataset/color"))  
  
print(len(os.listdir("plantvillage dataset/grayscale")))  
print(os.listdir("plantvillage dataset/grayscale"))  
  
[ ] ['segmented', 'grayscale', 'color']  
38  
['Cherry_(including_sour)_healthy', 'Peach_Bacterial_spot', 'Tomato_Tomato_mosaic_virus', 'Grape_Leaf_blight_(Isariopsis_Leaf_Spot)', 'Pepper_bell_healthy', 'Potato_Late_brown_necrosis']  
38  
['Cherry_(including_sour)_healthy', 'Peach_Bacterial_spot', 'Tomato_Tomato_mosaic_virus', 'Grape_Leaf_blight_(Isariopsis_Leaf_Spot)', 'Pepper_bell_healthy', 'Potato_Late_brown_necrosis']  
38  
['Cherry_(including_sour)_healthy', 'Peach_Bacterial_spot', 'Tomato_Tomato_mosaic_virus', 'Grape_Leaf_blight_(Isariopsis_Leaf_Spot)', 'Pepper_bell_healthy', 'Potato_Late_brown_necrosis']
```

The second cell contains code to load and display a leaf image:

```
(x) [ ] image_path = '/content/plantvillage_dataset/color/Apple_Cedar_apple_rust/025b2b9a-0ec4-4132-96ac-7f2832' 
```

3.Data Preprocessing using ImageDataGenerator

The screenshot shows a Jupyter Notebook cell containing code for setting up an `ImageDataGenerator`:

```
(x) [ ] datagen = ImageDataGenerator(rescale=1./255, rotation_range=20,  
    height_shift_range=0.2,  
    shear_range=0.2,  
    zoom_range=0.2,  
    horizontal_flip=True,  
    fill_mode='nearest', validation_split=0.2)  
  
train_generator = datagen.flow_from_directory(  
    dir,  
    target_size=(224, 224),  
    batch_size=32,  
    class_mode='categorical',  
    subset='training')  
test_generator = datagen.flow_from_directory(  
    dir,  
    target_size=(224, 224),  
    batch_size=32,  
    class_mode='categorical',  
    subset='validation')
```

Output from the cell:

```
[ ] Found 43456 images belonging to 38 classes.  
[ ] Found 10849 images belonging to 38 classes.
```

4.CNN Model Building

```

from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout
from tensorflow.keras.layers import BatchNormalization, ReLU

model=Sequential()
model.add(Conv2D(filters=32,kernel_size=3,padding='same',activation='relu',input_shape=[224,224,3]))

model.add(Conv2D(filters=32,kernel_size=3,padding='same',activation='relu'))
model.add(MaxPooling2D(pool_size=2,strides=2))
model.add(Conv2D(filters=64,kernel_size=3,padding='same',activation='relu'))
model.add(Conv2D(filters=64,kernel_size=3,padding='same',activation='relu'))
model.add(MaxPooling2D(pool_size=2,strides=2))

model.add(Flatten())
model.add(Dense(units=256,activation='relu'))
model.add(Dropout(0.25))
model.add(Dense(units=train_generator.num_classes,activation='softmax'))

```

5. Model Architecture/ Summary

```

model.add(Flatten())
model.add(Dense(units=256,activation='relu'))
model.add(Dropout(0.25))
model.add(Dense(units=train_generator.num_classes,activation='softmax'))

model.summary()

```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 224, 224, 3)	896
conv2d_1 (Conv2D)	(None, 224, 224, 3)	9,744
max_pooling2d (MaxPooling2D)	(None, 112, 112, 3)	0
conv2d_2 (Conv2D)	(None, 112, 112, 64)	18,496
conv2d_3 (Conv2D)	(None, 112, 112, 64)	16,928
max_pooling2d_1 (MaxPooling2D)	(None, 56, 56, 64)	0
flatten (Flatten)	(None, 30720)	0
dense (Dense)	(None, 256)	51,388,480
dropout (Dropout)	(None, 256)	0
dense_1 (Dense)	(None, 3)	9,766

6. Compiling and fitting the model

```

model.compile(optimizer='adam',loss='categorical_crossentropy',metrics=['accuracy'])

history=model.fit(train_generator,epochs=5,validation_data=test_generator)

```

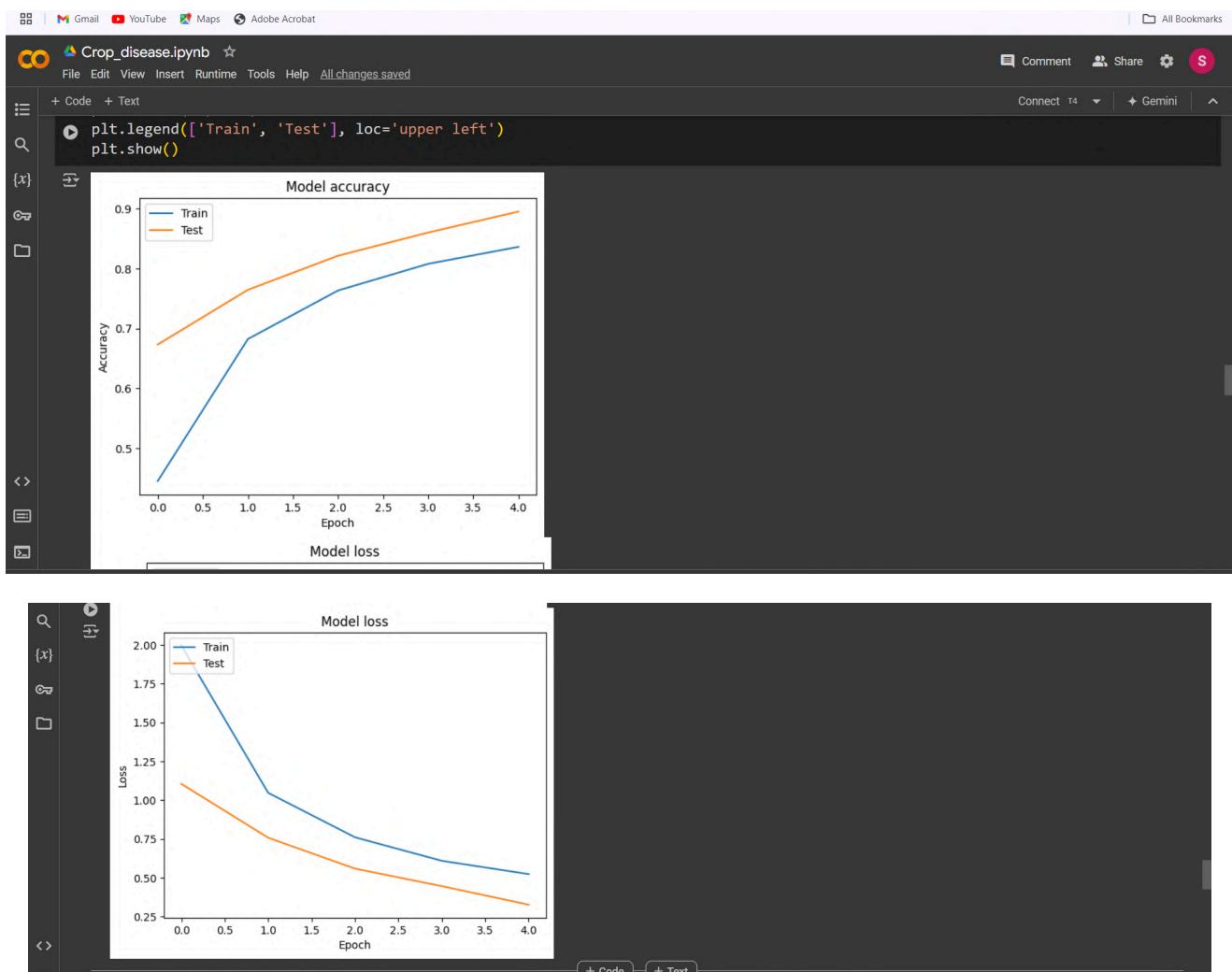
Epoch 1/5
 /usr/local/lib/python3.10/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset` class should call `super().__init__(**kwargs)` in its
 self._warn_if_super_not_called()
 1358/1358 703s 505ms/step - accuracy: 0.3251 - loss: 2.6437 - val_accuracy: 0.6731 - val_loss: 1.1051
 Epoch 2/5
 1358/1358 666s 488ms/step - accuracy: 0.6554 - loss: 1.1459 - val_accuracy: 0.7645 - val_loss: 0.7580
 Epoch 3/5
 1358/1358 675s 496ms/step - accuracy: 0.7534 - loss: 0.7900 - val_accuracy: 0.8214 - val_loss: 0.5590
 Epoch 4/5
 1358/1358 678s 497ms/step - accuracy: 0.7973 - loss: 0.6443 - val_accuracy: 0.8602 - val_loss: 0.4461
 Epoch 5/5
 1358/1358 676s 496ms/step - accuracy: 0.8293 - loss: 0.5379 - val_accuracy: 0.8951 - val_loss: 0.3267

```

plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['Train', 'Test'], loc='upper left')
plt.show()

```

7. Model accuracy and loss on train and test set



8. Test Accuracy is 89.59%

```
[1]: print("Evaluating model...")
val_loss, val_accuracy = model.evaluate(test_generator, steps=test_generator.samples // 32)
print(f"Validation Accuracy: {val_accuracy * 100:.2f}%")

[2]: print("Evaluating model...")
train_loss, train_accuracy = model.evaluate(train_generator, steps=train_generator.samples // 32)
print(f"Train Accuracy: {train_accuracy * 100:.2f}%")

[3]: def load_and_preprocess_image(image_path, target_size=(224, 224)):
    # Load the image
    img = Image.open(image_path)
    # Resize the image
    img = img.resize(target_size)
    # Convert the image to a numpy array
    img_array = np.array(img)
    # Add batch dimension
```

The screenshot shows a Jupyter Notebook cell containing Python code for evaluating a machine learning model. The code uses the `evaluate` method on a test data generator. The validation accuracy is printed as 89.59%. Below this, another cell evaluates the model on the training data, resulting in a train accuracy of 90.83%. At the bottom, a function `load_and_preprocess_image` is defined, which loads an image, resizes it, converts it to a numpy array, and adds a batch dimension.

9. Predicting a class of image and saving the model as .h5

The screenshot shows a Jupyter Notebook interface with the file 'Crop_disease.ipynb' open. The code cell contains three print statements. The first two print statements show validation and train accuracy metrics over multiple steps. The third cell contains a function definition for loading and preprocessing images.

```
[ ] print("Evaluating model...")
val_loss, val_accuracy = model.evaluate(test_generator, steps=test_generator.samples // 32)
print(f"Validation Accuracy: {val_accuracy * 100:.2f}%")

[ ] Evaluating model...
339/339 129s 380ms/step - accuracy: 0.8962 - loss: 0.3345
Validation Accuracy: 89.59%


[ ] print("Evaluating model...")
train_loss, train_accuracy = model.evaluate(train_generator, steps=train_generator.samples // 32)
print(f"Train Accuracy: {train_accuracy * 100:.2f}%")

[ ] Evaluating model...
1358/1358 512s 377ms/step - accuracy: 0.9082 - loss: 0.2936
Train Accuracy: 90.83%


[ ] def load_and_preprocess_image(image_path, target_size=(224, 224)):
    # Load the image
    img = Image.open(image_path)
    # Resize the image
    img = img.resize(target_size)
    # Convert the image to a numpy array
    img_array = np.array(img)
    # Add batch dimension
```

14 . Conclusion

Farmologist is an innovative app designed to empower farmers by leveraging AI and data-driven insights to enhance agricultural productivity and sustainability. By integrating features such as crop health monitoring, weather forecasting, and financial services, Farmologist provides farmers with a comprehensive toolset that simplifies decision-making and optimizes farming practices. The app's user-friendly interface ensures accessibility for farmers, even in rural areas, enabling them to access real-time market prices, apply for loans, and receive personalized farming recommendations. As it continues to evolve, Farmologist aims to bridge the gap between traditional farming methods and modern technology, ultimately helping farmers achieve greater efficiency, profitability, and resilience in the face of ever-changing agricultural challenges.

15 . References

Few Research Papers-

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