

AgriLearn (AI Platform)

Course Name: GENRATIVE AI

Institution Name: Medicaps University – Datagami Skill Based Course

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Problem Statement & Objectives

1. Problem Statement

Agriculture remains one of the most critical sectors for economic development, food security, and sustainable growth. However, agricultural education often relies on static learning materials, traditional classroom instruction, and standardized assessments that do not adapt to individual learning needs. Students studying agriculture frequently encounter challenges in understanding practical farming concepts such as soil health management, irrigation planning, crop nutrition, pest control, and sustainable agricultural practices.

Conventional assessment systems primarily focus on scoring mechanisms rather than conceptual reinforcement. Learners are typically provided with marks or correct answers without detailed explanations that clarify *why* a particular answer is correct or incorrect. This approach limits deep learning and does not help students identify their knowledge gaps. As a result, many learners memorize information without fully understanding real-world agricultural applications.

Additionally, the absence of personalized feedback prevents learners from improving systematically. In agriculture, where decisions directly impact productivity, sustainability, and environmental balance, conceptual clarity is essential. Therefore, there is a strong need for an intelligent, AI-driven educational system that can dynamically generate assessments, analyze learner responses, provide contextual explanations, and create a continuous improvement learning loop.

The AgriLearn AI platform addresses this need by integrating Generative AI to transform traditional agricultural assessments into an adaptive, interactive, and concept-reinforcing educational experience.

2. Project Objectives

The primary objective of the AgriLearn AI platform is to develop an interactive AI-powered educational bot tailored for the Agriculture domain. The system aims to enhance conceptual understanding through intelligent assessment and personalized evaluation.

The key objectives of the project are:

1. To develop a dynamic quiz generation system

The platform should generate topic-based multiple-choice questions using a Large Language Model instead of relying on static question banks.

2. To implement AI-driven contextual evaluation

The system should analyze user responses and provide deep explanations for incorrect answers, helping learners understand conceptual mistakes.

3. To create a personalized learning loop

The platform should identify knowledge gaps and reinforce domain-specific agricultural concepts through structured feedback.

4. To ensure secure and stateless architecture

The application must maintain zero data retention, manage session states efficiently, and securely integrate external AI APIs.

5. To optimize performance using model-tiering

The system should use faster AI models for question generation and reasoning-intensive models for detailed evaluation.

6. To promote adaptive and scalable learning

The application should be designed in a way that allows future expansion into additional agricultural domains and scalable deployment environments.

3. Scope of the Project

The scope of the AgriLearn AI platform includes the design and implementation of an AI-powered web-based assessment tool specifically for agriculture-related learning.

In Scope:

- Generation of farming-related MCQs using Generative AI.
- Topic and difficulty-based quiz customization.
- Structured JSON-based question parsing.
- Real-time evaluation and contextual feedback generation.
- Session-based state management using Streamlit.
- Secure API key management and HTTPS communication.
- Stateless and scalable architecture design.

Out of Scope:

- Persistent database storage of user data.
- Long-term user performance tracking.
- Infrastructure provisioning (server deployment setup).
- Mobile application development.
- Integration with external agricultural datasets.

The project is primarily focused on demonstrating the integration of Generative AI in domain-specific education and showcasing intelligent assessment systems rather than building a full-scale enterprise-level learning management system.

Proposed Solution

1. Key Features

- The AgriLearn AI platform incorporates the following major features:
- Dynamic AI-generated agriculture-based multiple-choice questions
- Topic and difficulty level selection

- Strict JSON-structured response enforcement
- AI-powered contextual evaluation
- Personalized feedback mechanism
- Session-based state management
- Secure API key handling using secrets management
- Stateless and scalable system architecture
- Model-tiering for performance optimization
- Zero data retention policy

2. Overall Architecture / Workflow

The AgriLearn AI platform follows a monolithic serverless frontend architecture, where the entire application logic/presentation layer are handled using Streamlit. The system integrates externally with Google Generative AI models through secure API communication.

The workflow of the system is divided into three structured phases:

Phase 1: Setup Phase

In this stage, the user selects:

- Agriculture topic (e.g., Soil Health, Irrigation Management, Sustainable Farming, Crop Nutrition)
- Difficulty level (Beginner, Intermediate, Advanced)

These parameters are captured via the sidebar interface and formatted into a structured prompt. The application then prepares a request for AI-based question generation.

Phase 2: Assessment Phase

1. The formatted prompt is sent to the Generative AI model.
2. The AI returns a strict JSON array containing:
 - Question
 - Four options
 - Correct answer
 - Hint
3. The application parses the JSON output into Python dictionaries.
4. Questions are stored temporarily in `st.session_state`.
5. The user answers questions sequentially using radio buttons.
6. Progress tracking ensures structured navigation through the quiz.

Phase 3: Evaluation Phase

Once all responses are submitted:

1. The system compiles:
 - Questions
 - Correct answers
 - User-selected answers
2. A second AI prompt is generated for deep analysis.
3. A reasoning-heavy model evaluates:
 - Incorrect selections
 - Conceptual misunderstandings
 - Knowledge gaps
4. The AI returns structured Markdown feedback.
5. The final results display:
 - Score
 - Performance summary
 - Detailed contextual explanations

This transforms the system from a simple quiz engine into an intelligent agriculture tutor.

Architectural Characteristics

- **Stateless between sessions** – No persistent database storage.
 - **Session-based state machine** – Controls application flow (setup → quiz → results).
 - **Synchronous execution handling** – Uses controlled reruns to prevent redundant API calls.
 - **Secure HTTPS communication** – Ensures encrypted data transmission.
 - **Prompt schema enforcement** – Prevents malformed AI outputs.
- The architecture ensures scalability, privacy, and efficient performance.

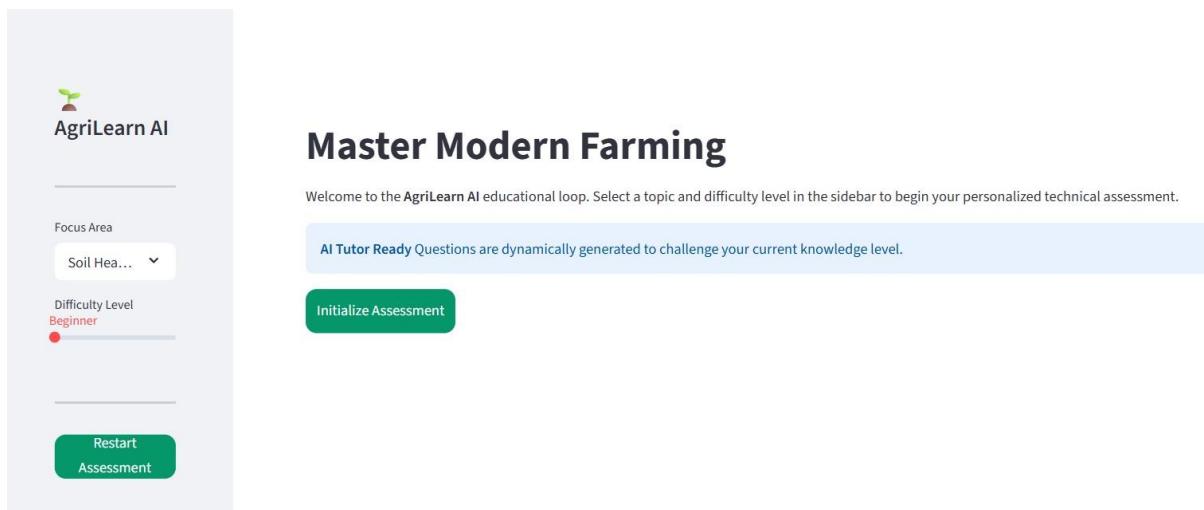
Results & Output

1. Screenshots / Outputs

The AgriLearn AI platform was successfully implemented and tested in a Streamlit-based web environment. The application produces dynamic and interactive outputs at each stage of the workflow.

- Sidebar-based topic selection (e.g., Soil Health, Irrigation, Crop Nutrition, Sustainable Farming).
- Difficulty level selection (Beginner, Intermediate, Advanced).
- “Start Assessment” trigger button.
- Clean and responsive UI layout.

This stage ensures that the system dynamically customizes quiz generation based on user input.



Assessment Phase Output

- AI-generated multiple-choice questions (5 per session).
- Four structured options per question.
- Progress indicator showing quiz completion status.
- Hint system to assist learners.
- Sequential navigation through questions.

Each quiz is uniquely generated using a Large Language Model, ensuring variability and freshness of content. The system successfully parses the strict JSON response from the AI without errors, demonstrating reliable schema enforcement.

Question 3 of 5 — Sustainable Irrigation

Which of the following best describes the 'Crop Coefficient' (Kc) used in irrigation scheduling?

Select the best answer:

- The ratio of the actual crop evapotranspiration to the reference evapotranspiration
- The percentage of nitrogen required per unit of water applied
- The maximum depth to which a specific crop's roots can grow
- The rate at which water infiltrates a specific soil type

Pro Tip: It is a multiplier that adjusts the standard ET₀ value based on the specific crop type and its current growth stage.

Submit Answer

After submission of all responses, the system generates:

- Total score and percentage.
- Question-wise performance analysis.
- Detailed explanation for each incorrect answer.
- Concept clarification and reinforcement suggestions.
- Structured Markdown-based feedback display.

Unlike traditional quiz systems that only provide correct answers, the AgriLearn AI platform delivers contextual explanations that help users understand the scientific reasoning behind correct farming practices.

Assessment Analysis

Expertise Score
20.0%
+ 1/5 Correct

Deep Contextual Feedback

As an expert agricultural scientist, I have reviewed your responses to the irrigation science questions. This evaluation provides a deep contextual analysis, elucidates the underlying scientific principles, and addresses potential misconceptions inherent in your selections.

Expert Evaluation of Irrigation Science Assessment

Your performance on this assessment demonstrates a foundational understanding in some areas of sustainable irrigation, alongside specific conceptual gaps that are critical in advanced agricultural water management. Let's delve into each response.

1. Soil Water Tension Measurement for Precision Irrigation

Q: Which device is primarily used in sustainable irrigation to measure soil water tension, allowing for precise scheduling based on plant needs? User: Tensiometer Correct: Tensiometer

Evaluation: **Correct.** Your answer is spot-on and indicates a sound understanding of a fundamental tool in precision irrigation.

Scientific Explanation: A tensiometer is a cornerstone instrument for direct measurement of soil water potential, often expressed as soil water tension or suction. It consists of a porous ceramic cup connected via a water-filled tube to a vacuum gauge. As the soil dries, water is drawn out. The brilliance of the tensiometer lies in its direct correlation to plant water availability. Plants expend energy to extract water from the soil; the higher the soil water tension, the more energy required. By monitoring soil water tension, growers can determine precisely when irrigation is needed.

2. Goals of Regulated Deficit Irrigation (RDI)

Q: In the context of Regulated Deficit Irrigation (RDI), what is the primary goal during the non-critical growth stages of a crop? User: To maximize vegetative biomass production Correct: To reduce water application to induce beneficial stress and save resources

Evaluation: **Incorrect.** Your response reflects a common misconception regarding the core principles and objectives of RDI.

3. Key Outcomes

The AgriLearn AI platform achieved the following outcomes:

1. Successfully implemented dynamic AI-based quiz generation without static question banks.
2. Achieved reliable JSON schema enforcement for structured AI output.
3. Delivered contextual, reasoning-based explanations rather than simple scoring.
4. Maintained zero data retention to ensure privacy and security.
5. Optimized performance by separating question generation and evaluation into different model tiers.
6. Prevented redundant API calls using session-based caching.
7. Demonstrated scalability due to stateless architecture.

The project successfully bridges the gap between theoretical agricultural education and practical conceptual understanding by leveraging Generative AI

Conclusion

The AgriLearn AI platform successfully demonstrates the practical application of Generative AI in transforming domain-specific education, particularly within the Agriculture sector. The project addresses a critical limitation of traditional agricultural learning systems — the lack of personalized feedback and conceptual reinforcement.

By integrating Large Language Models into an interactive assessment system, the platform moves beyond static quizzes and simple scoring mechanisms.

Instead of merely indicating correct or incorrect answers, the system analyzes learner responses, identifies conceptual gaps, and provides detailed contextual explanations. This approach promotes deeper understanding of farming practices such as soil management, irrigation strategies, crop nutrition, pest control, and sustainable agricultural techniques.

Technically, the project showcases strong implementation of:

- AI-based dynamic content generation
- JSON schema enforcement for structured outputs
- Secure API integration
- Session-based finite state management
- Stateless architecture for scalability
- Model-tiering for performance optimization

The use of Streamlit enabled the development of a responsive and interactive user interface, while the integration of Google Generative AI models ensured intelligent and adaptive evaluation capabilities. The implementation of zero data retention policies further strengthened privacy and security considerations.

From an academic and technical perspective, the project provided valuable experience in:

- Prompt engineering and LLM interaction
- Managing synchronous application flows
- Preventing redundant API calls
- Designing scalable AI-driven systems
- Building adaptive learning applications

Overall, AgriLearn AI successfully bridges the gap between theoretical agricultural knowledge and real-world application. The platform proves that Generative AI can significantly enhance personalized learning experiences and has the potential to revolutionize domain-specific education systems.

Future Scope & Enhancements

While the AgriLearn AI platform successfully demonstrates an intelligent AI-driven agricultural assessment system, there are several opportunities for further enhancement and expansion. Future improvements can significantly increase the platform's impact, scalability, and practical relevance.

1. User Authentication & Performance Tracking

Currently, the system follows a zero-retention stateless architecture. In future versions, user authentication can be integrated to:

- Track individual learning progress over time
- Store historical performance data
- Provide personalized performance analytics
- Generate long-term improvement reports

This would transform the system into a more structured learning management platform.

2. Adaptive Difficulty Adjustment

An enhancement can be implemented where the system dynamically adjusts question difficulty based on user performance. For example:

- If a user performs well, the system increases complexity.
- If a user struggles, the system provides easier reinforcement questions.

This would create a fully adaptive learning ecosystem powered by AI.

3. Multilingual Support

Since agriculture is highly relevant in rural and regional areas, the platform can be expanded to support multiple languages such as:

- Hindi
- Regional Indian languages
- Other global languages

This would increase accessibility and usability among farmers and agriculture students from diverse backgrounds.

4. Integration of Multimedia Learning

Future versions can include:

- Image-based questions (e.g., crop disease identification)
- Infographics for irrigation systems
- Short explanatory videos
- Diagram-based problem-solving

This would improve visual learning and practical understanding.

5. Real-Time Agricultural Data Integration

The system can be enhanced by integrating:

- Weather APIs

- Soil health datasets

- Market price data
- Government agricultural schemes

This would allow the bot to provide real-world, real-time contextual learning instead of purely theoretical assessments.

6. Advanced Analytics Dashboard

An administrative dashboard can be developed for:

- Institutional usage
- Performance trend analysis
- Topic-wise weakness identification
- Aggregate learning statistics

This would make the platform useful for universities and training institutes.

7. Mobile & Offline Accessibility

The platform can be optimized into:

- A Progressive Web Application (PWA)
- Android/iOS mobile application
- Offline quiz mode with limited AI caching

This would increase reach among rural learners with limited internet access.

8. Expansion to Broader Agricultural Domains

The platform can expand into specialized modules such as:

- Precision Agriculture
- Agri-Economics
- Agri-Entrepreneurship

- Sustainable Climate-Resilient Farming

This would make the system industry-ready and aligned with modern agricultural advancements.

Final Outlook

With these enhancements, AgriLearn AI can evolve from a smart assessment tool into a comprehensive AI-powered agricultural learning ecosystem. The integration of adaptive intelligence, multilingual accessibility, real-time datasets, and performance analytics can significantly improve agricultural education and skill development at scale.