Problem Statement No.9

Al Agent for Smart Farming Advice

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OUTLINE

- Problem Statement
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PROBLEM STATEMENT

Example: The Challenge – An AI Agent for Smart Farming Advice, powered by RAG (Retrieval-Augmented Generation), supports small-scale farmers by delivering real-time, localized agricultural guidance. It retrieves trusted data on weather forecasts, soil conditions, crop recommendations, pest control measures, and current market prices from agricultural departments, meteorological sources, and agritech platforms.

Farmers can interact in their local language and ask questions like "What crop is best for this season?" or "What is today's mandi rate for tomatoes?"

The agent ensures timely, data-driven decisions that reduce risk, increase yield, and boost income.

This AI-driven assistant bridges the knowledge gap and brings smart farming to the grassroots.



PROPOSED SOLUTION

We propose a smart AI Agent, built using Retrieval-Augmented Generation (RAG), that provides localized and actionable farming advice to small-scale farmers.

Key Features:

Natural Language Understanding – Accepts user queries in plain language (e.g., "What crop should I grow this month?").

Document Retrieval – Pulls data from trusted sources like:

- Weather APIs
- Soil condition databases
- Agricultural research documents
- Market price feeds

Contextual Response Generation – Uses IBM Granite LLM to generate accurate and helpful answers.



SYSTEM APPROACH

Technologies Used:

- **IBM Watsonx.ai / Granite LLM:** For language generation
- **IBM Cloud Functions:** For serverless API execution
- **IBM Cloudant / DB2:** For structured data storage
- **Python (Flask):** Backend development
- LangChain or Haystack: RAG pipeline integration
- **HTML + JavaScript:** Frontend chatbot interface

RAG Architecture Flow:

• Query from user \rightarrow Document Retrieval (weather, soil, etc.) \rightarrow Prompt formatting \rightarrow LLM generation \rightarrow Response to user



ALGORITHM & DEPLOYMENT

Core Algorithm: Retrieval-Augmented Generation (RAG)

RAG is a hybrid framework that retrieves external knowledge and combines it with large language model (LLM) capabilities to generate intelligent, grounded answers.

How It Works:

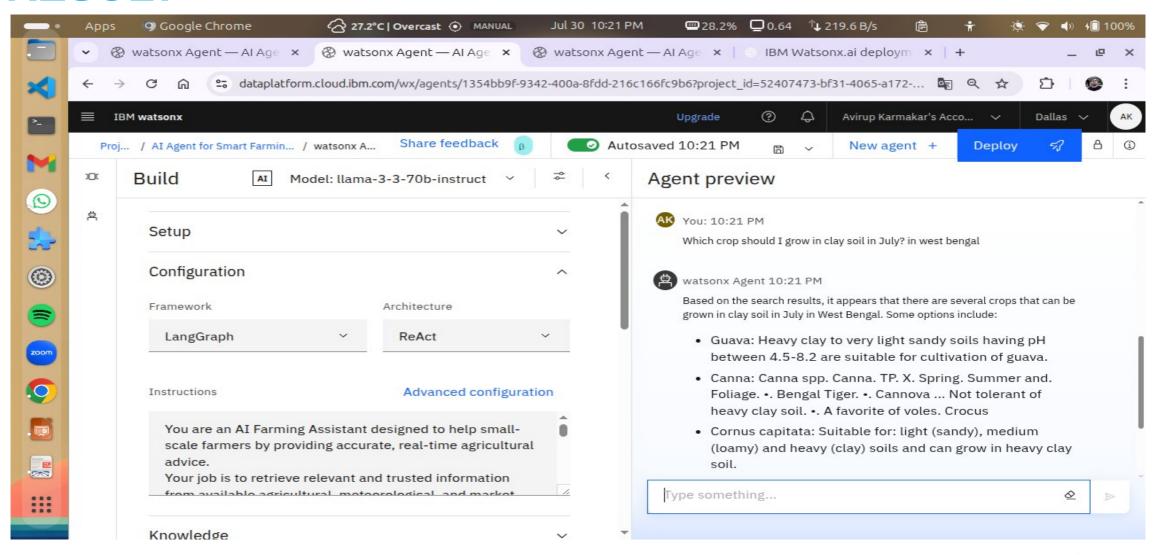
- User Query Input: Natural language query from farmer (e.g., "Best crop for loamy soil in July?")
- Vector Similarity Search:
 - The query is encoded into a vector using sentence embeddings.
 - The system searches a vector database (e.g., FAISS) to find semantically similar documents (weather, soil data, crop info).

Document Retrieval:

- Top relevant documents are retrieved and ranked.
- These may include recent weather forecasts, soil analysis, and government agricultural advisories.
- Context Fusion: Retrieved documents + the original query are combined into a single prompt.
- LLM Response Generation:
 - Prompt is passed to IBM Granite (LLM via Watsonx.ai).
 - The model generates a response based on real data, ensuring both fluency and accuracy.

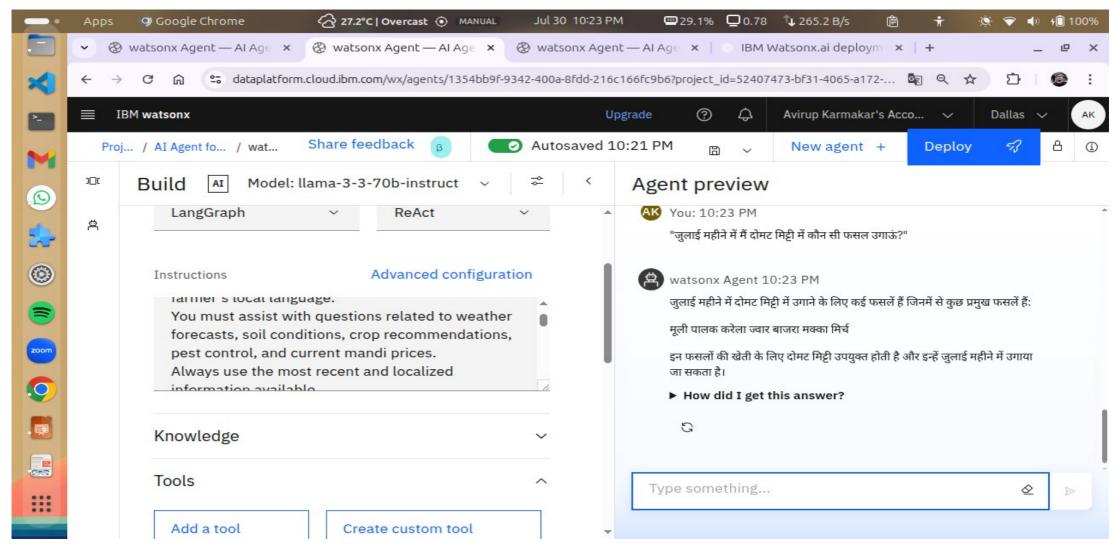


RESULT



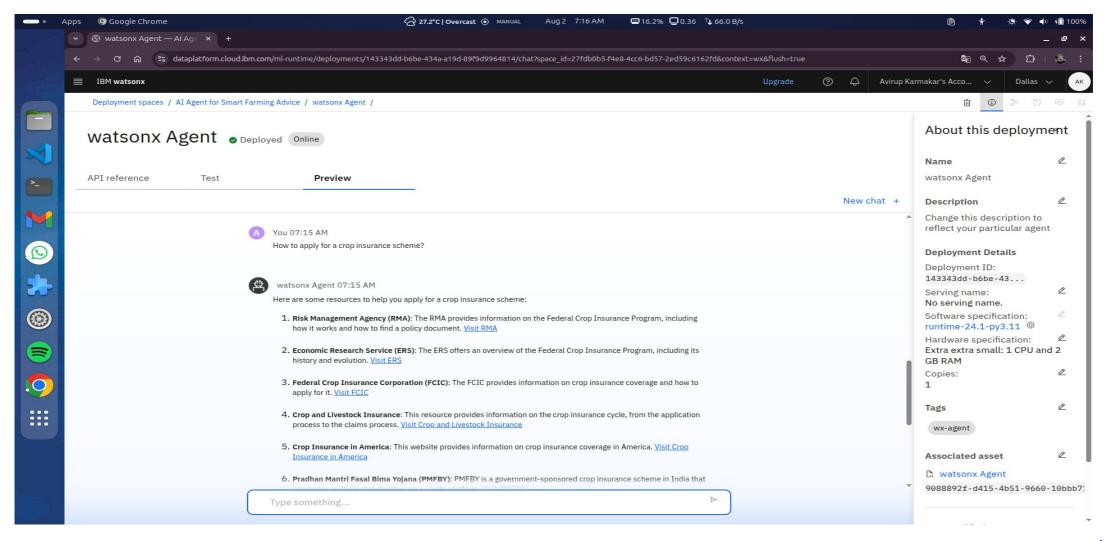


RESULT





RESULT





CONCLUSION

The AI agent empowers farmers by enabling data-driven decision-making through real-time, localized agricultural insights. It enhances access to expert-level knowledge, even in remote or underserved regions, by delivering context-aware responses in natural language. The system's design is scalable and adaptable to various languages and geographies, making it suitable for widespread use. By integrating Retrieval-Augmented Generation (RAG) with large language models on IBM Cloud, the solution achieves a balance of accuracy, affordability, and scalability, offering a practical path forward for smart and sustainable farming.



FUTURE SCOPE

- The AI agent can be further enhanced to serve a broader range of farmers and agricultural needs. Future improvements include:
 - Expanding to multiple Indian languages to ensure accessibility across diverse regions;
 - Integrating with IoT soil sensors for real-time, data-driven crop recommendations;
 - Using drone imagery to assess crop health and detect diseases early;
 - Adding a voice-based assistant to support illiterate or elderly users; and
 - Implementing real-time alerts for pest outbreaks, extreme weather, or market price changes.
- These advancements will make the system smarter, more inclusive, and aligned with the goals of precision and sustainable farming.



REFERENCES

- IBM Watsonx.ai Documentation
- IBM Cloud Lite Functions
- LangChain RAG Framework
- Kaggle, Indian Meteorological Dept. for weather APIs
- Government agriculture portal (e.g., https://mkisan.gov.in)



IBM CERTIFICATIONS

Screenshot/ credly certificate(getting started with AI)

Getting Started with In recognition of the commitment to achieve Artificial Intelligence professional excellence Avirup Karmakar Has successfully satisfied the requirements for: Getting Started with Artificial Intelligence Issued on: Jul 16, 2025 Issued by: IBM SkillsBuild Verify: https://www.credly.com/badges/9812e968-19fa-4949-8c7c-89c8ab34e1e0



IBM CERTIFICATIONS

Screenshot/ credly certificate(Journey to Cloud)

In recognition of the commitment to achieve professional excellence Avirup Karmakar Has successfully satisfied the requirements for: Journey to Cloud: Envisioning Your Solution Issued on: Jul 17, 2025 Issued by: IBM SkillsBuild Verify: https://www.credly.com/badges/e56e6f8f-c81f-4d10-afd5-9055e3548864



IBM CERTIFICATIONS

Screenshot/ credly certificate(RAG Lab)

IBM SkillsBuild Completion Certificate



This certificate is presented to

Avirup Karmakar

for the completion of

Lab: Retrieval Augmented Generation with LangChain

Completion date: 17 Jul 2025 (GMT)

Learning hours: 20 mins



THANK YOU

