
CAPSTONE PROJECT

PROJECT TITLE

Presented By:

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OUTLINE

- **Problem Statement** (Should not include solution)
- **Proposed System/Solution**
- **System Development Approach** (Technology Used)
- **Algorithm & Deployment**
- **Result (Output Image)**
- **Conclusion**
- **Future Scope**
- **References**

PROBLEM STATEMENT

Small-scale farmers in rural areas face challenges in accessing timely and accurate agricultural information. Limited exposure to advanced farming techniques, lack of awareness of real-time market trends, unpredictable weather, and inadequate pest control measures hinder optimal crop production. Language barriers further isolate these communities from accessing vital resources. The need for hyper-localized, easily understandable agricultural advice remains unmet for many farmers, reducing productivity and income levels.

PROPOSED SOLUTION

We propose an AI-powered assistant for smart farming advice using Retrieval-Augmented Generation (RAG) integrated with IBM Cloud Lite and IBM Granite. This virtual agent retrieves data from trusted sources such as meteorological APIs, agricultural departments, and market databases to offer localized, real-time guidance in the user's native language. It provides recommendations on crop selection, pest control, soil health, weather updates, and mandi (market) prices. The system bridges the knowledge gap by offering data-driven, contextual answers for better decision-making.

SYSTEM APPROACH

- **IBM Granite Models:** Used for multilingual natural language understanding and generation.
- **IBM Watson Assistant (Lite):** For building the conversational interface.
- **RAG Architecture:** Combines document retrieval with LLM-based response generation.
- **IBM Cloud Object Storage:** For storing knowledge base documents (weather, market data, crop reports).

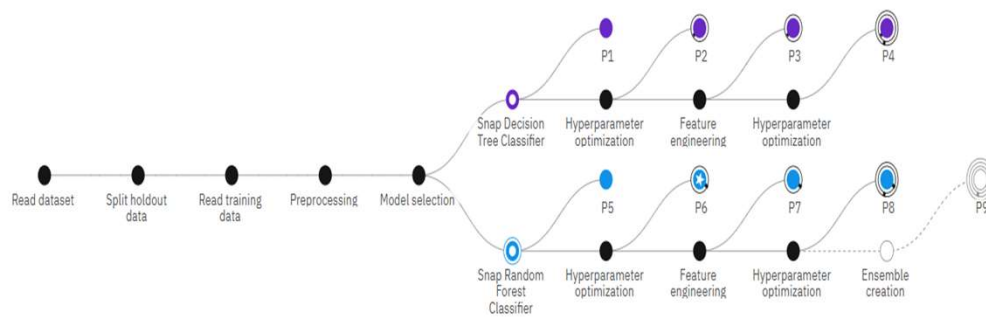
ALGORITHM & DEPLOYMENT

- **User Query Input:** User asks a question in their local language (e.g., “What crop is best for this soil?”).
- **Retriever Module:** Fetches relevant documents (from cloud storage, APIs).
- **Generator Module:** Uses IBM Granite to form a contextual and natural response.
- **Response Translation (if needed):** Translated back to the local language for display.
- **Deployment:**
 - Hosted on IBM Cloud Lite
 - Connected via **Watson Assistant**, exposed through a web/mobile interface.

RESULT

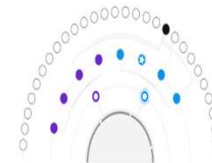
Progress map


Prediction column: label



Relationship map

[Swap view](#) 



Experiment completed 

8 PIPELINES GENERATED

8 pipelines generated from algorithms. See pipeline leaderboard below for more detail.

Time elapsed: 3 minutes

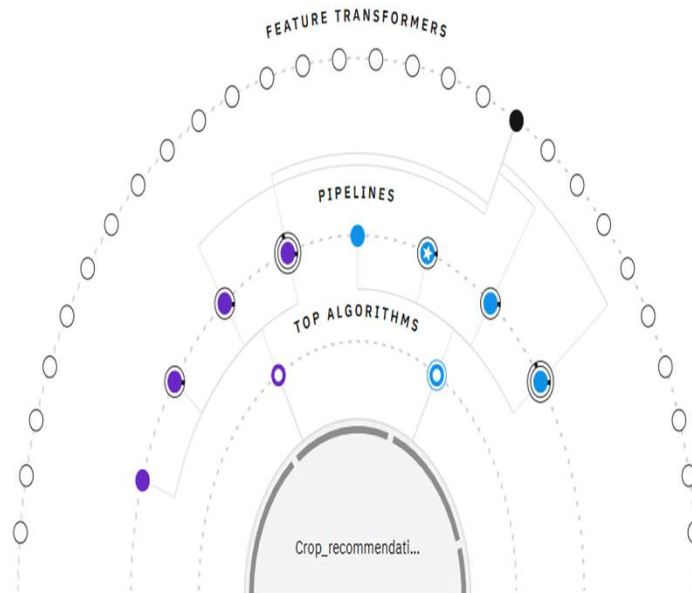
[View log](#)

[Save code](#)

RESULT

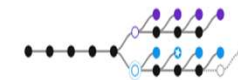
Relationship map ①

Prediction column: label



Progress map

[Swap view ↔](#)



Experiment completed 🟢

8 PIPELINES GENERATED

8 pipelines generated from algorithms. See pipeline leaderboard below for more detail.









Time elapsed: 3 minutes

[View log](#)

[Save code](#)

RESULT

Pipeline leaderboard ?

	Rank	↑	Name	Algorithm	Specialization	Accuracy (Optimized) Cross Validation	Enhancements	Build time
★	1		Pipeline 6	 Snap Random Forest Classifier		0.981	HPO-1	00:00:06
	2		Pipeline 5	 Snap Random Forest Classifier		0.981	None	00:00:01
	3		Pipeline 2	 Snap Decision Tree Classifier		0.980	HPO-1	00:00:06
	4		Pipeline 1	 Snap Decision Tree Classifier		0.980	None	00:00:03
	5		Pipeline 8	 Snap Random Forest Classifier		0.979	HPO-1 FE HPO-2	00:00:37
	6		Pipeline 7	 Snap Random Forest Classifier		0.979	HPO-1 FE	00:00:29
	7		Pipeline 4	 Snap Decision Tree Classifier		0.976	HPO-1 FE HPO-2	00:00:38
	8		Pipeline 3	 Snap Decision Tree Classifier		0.976	HPO-1 FE	00:00:34

RESULT

Crop Deployment ✓ Deployed Online

API reference

Test

Enter input data

Text

JSON

Enter data manually or use a CSV file to populate the spreadsheet. Max file size is 50 MB.

[Download CSV template](#) ⬇

[Browse local files](#) ↗

[Search in space](#) ↗

[Clear all](#) ×

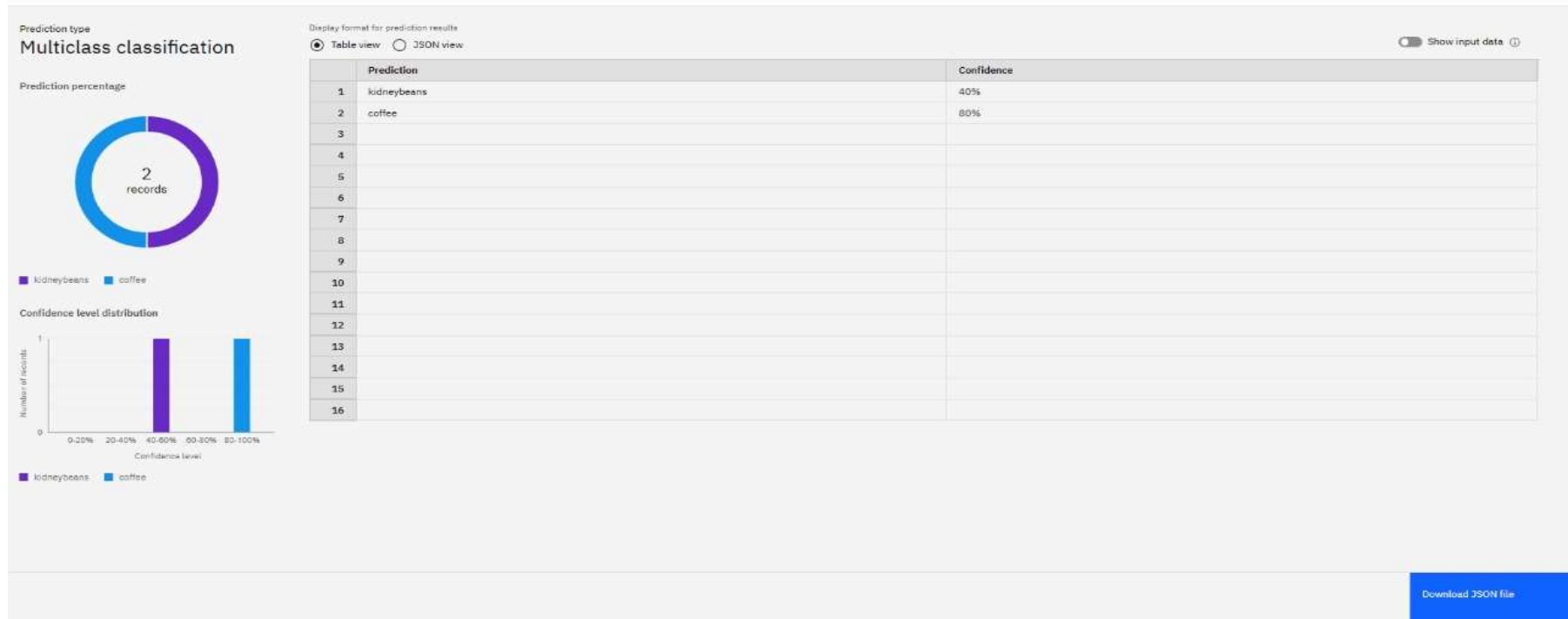
	N (double)	P (double)	K (double)	temperature (double)	humidity (double)	ph (double)	rainfall (double)
1	60	48	39	24.28	80.3	7.04	231
2	76	50	45	29	50	6	300
3							
4							
5							

2 rows, 7 columns

Predict

RESULT

Prediction results



CONCLUSION

- The AI agent effectively empowers small-scale farmers by delivering smart, localized agricultural advice in real time. Through the use of IBM's powerful AI and cloud technologies, farmers can make data-driven decisions on crop planning, market timing, and pest management. The assistant enhances accessibility with multilingual interaction, offering an intuitive and impactful solution. This project showcases how RAG models combined with cloud services can address grassroots-level challenges in agriculture.

FUTURE SCOPE

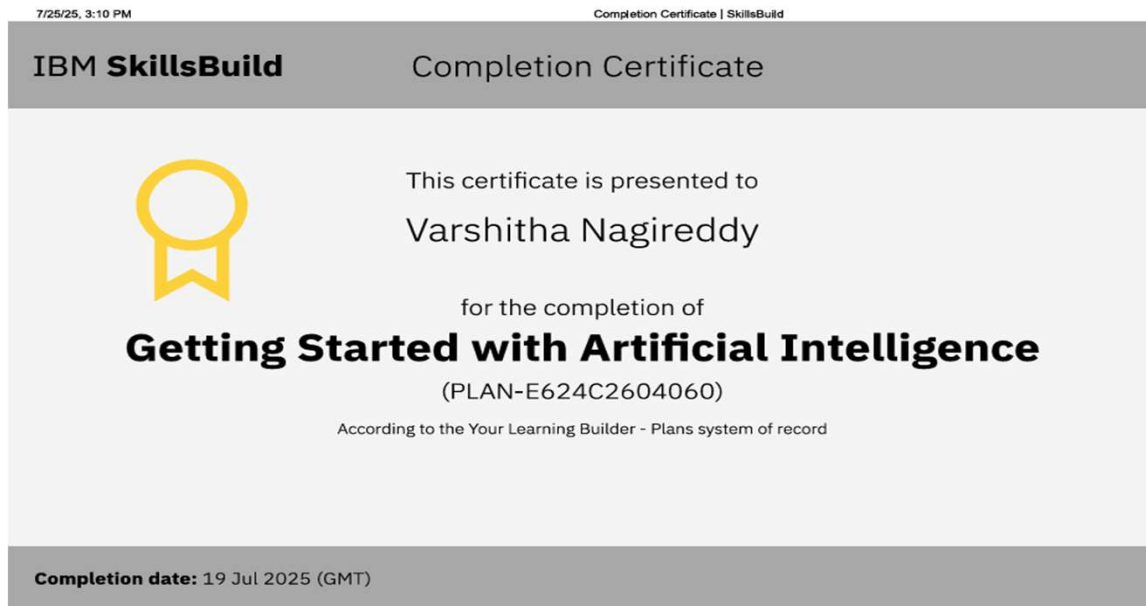
- Voice Assistant Integration for accessibility among non-literate farmers.
- IoT Integration to collect real-time soil and weather data from sensors.
- Offline Mode with Syncing for areas with limited internet.
- Drone Surveillance Support for pest monitoring and precision agriculture.
- Blockchain Integration for supply chain traceability and secure farmer payments.

REFERENCES

1. IBM Granite Documentation – <https://www.ibm.com/products/granite>
2. IBM Watson Assistant – <https://cloud.ibm.com/catalog/services/watson-assistant>
3. IBM Cloud Lite – <https://cloud.ibm.com>
4. Indian Meteorological Department APIs
5. Dataset: Crop_recommendation.csv

IBM CERTIFICATIONS

- Screenshot/ credly certificate(getting started with AI)



IBM CERTIFICATIONS

- Screenshot/ credly certificate(Journey to Cloud)



IBM CERTIFICATIONS

- Screenshot/ credly certificate(RAG Lab)



THANK YOU