# CAPSTONE PROJECT RECIPE PREPARATION AGENT

#### **Presented By:**

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#### **OUTLINE**

- Problem Statement
- Proposed Solution
- System Development Approach (Technology Used)
- Algorithm & Deployment
- Result (Output Image)
- Conclusion
- Future Scope
- References



# PROBLEM STATEMENT

#### Statement 16.

The **Recipe Preparation Agent** helps users cook meals using only the ingredients they have on hand. By inputting available groceries, users receive tailored recipe suggestions using a RAG-based AI system. The agent retrieves relevant recipes and generates step-by-step instructions adapted to ingredient limitations. It offers substitutions, cooking tips, and dietary adjustments based on user preferences or restrictions. Designed to reduce food waste and save time, it turns pantry items into practical meal solutions.

This AI assistant makes everyday cooking smarter, simpler, and more sustainable.



# PROPOSED SOLUTION

The proposed system addresses the challenge of helping users cook meals based on the limited ingredients they have on hand. By leveraging IBM Cloud's Watsonx, Granite LLMs, and Agent Lab, the solution uses advanced AI techniques to retrieve and generate custom recipe suggestions while minimizing food waste and optimizing user experience. The solution will consist of the following components:

#### Data Collection

- A dataset of ~13,000 structured recipes by Joseph R. Martinez is uploaded to IBM Cloud Object Storage.
- Each recipe contains ingredients, instructions, cuisine tags, and dietary details.

#### Retrieval Mechanism (RAG-based)

- The system uses a Retrieval-Augmented Generation (RAG) approach to fetch relevant recipes based on the user's input ingredients.
- Recipes are retrieved using semantic similarity from the stored dataset.

#### Al-Powered Adaptation using Watsonx Granite Models

- The retrieved recipe is passed to IBM's Granite foundation model to adapt the instructions based on ingredient availability.
- Cooking tips and modifications are suggested according to the user's dietary preferences.

#### Agentic Workflow (Agent Lab)

- The entire logic is implemented as an AI agent within Watsonx Agent Lab.
- The agent combines retrieval, prompt engineering, and model inference in a single workflow.

#### Deployment

- The Al agent is deployed through Watsonx Runtime and accessed via the Endpoint URL.
- This ensures scalability and seamless user interaction.

#### Evaluation

- The system is tested based on user satisfaction, relevance of suggested recipes, and response accuracy.
- Future improvements include user feedback integration for recipe quality and personalization tuning.



# SYSTEM APPROACH

The "System Approach" section outlines the overall strategy and methodology for developing and implementing the Recipe Preparation Agent using IBM Cloud services. The system leverages AI and RAG-based techniques integrated via Agent Lab to deliver customized recipe recommendations. Here's the structure for this section:

#### System Requirements

- IBM Cloud account
- Watsonx.ai access with Agent Lab enabled
- IBM Granite 8-8-3b-instruct model
- IBM Cloud Object Storage (for uploading the recipe dataset)
- Recipe Dataset: ~13,000 structured recipes

#### Libraries & Tools Used

- IBM Watsonx Agent Lab (for creating and deploying agents)
- Watsonx Runtime (for executing LLMs via endpoint)
- IBM Granite LLM APIs (for recipe adaptation and text generation)
- Prompt Engineering Techniques (used within agent tools for tailoring outputs)



# **ALGORITHM & DEPLOYMENT**

This section outlines the algorithmic foundation and deployment strategy for the Recipe Preparation Agent. The solution utilizes a **Retrieval-Augmented Generation (RAG)** approach integrated with IBM's **Granite large language models**, enabling personalized recipe generation based on user-provided ingredients.

#### Algorithm Selection:

- The system uses a RAG (Retrieval-Augmented Generation) architecture. This approach combines:
- Retrieval of relevant recipe documents from a dataset stored in IBM Cloud Object Storage
- Generation of custom recipe instructions using the IBM Granite LLM hosted on Watsonx
- RAG is chosen for its strength in handling open-ended user queries by grounding generation on reliable, retrieved documents — ideal for cooking scenarios with dynamic ingredients and preferences.

#### Data Input:

The input to the system includes:

- A list of available ingredients entered by the user
- Optional preferences such as dietary restrictions or cuisine type
- These inputs are passed to the Watsonx agent, which uses them to guide the retrieval and generation process.



#### **GENERATION PROCESS AND DEPLOYMENT**

#### Generation Process

- Retrieved recipes are fed into IBM's Granite foundation model via the Watsonx Runtime
- The model adapts instructions to:
  - Remove unavailable ingredients
  - Suggest substitutions
  - Offer dietary/cooking tips
- The generation process is controlled via **prompt engineering** inside the Agent Lab environment

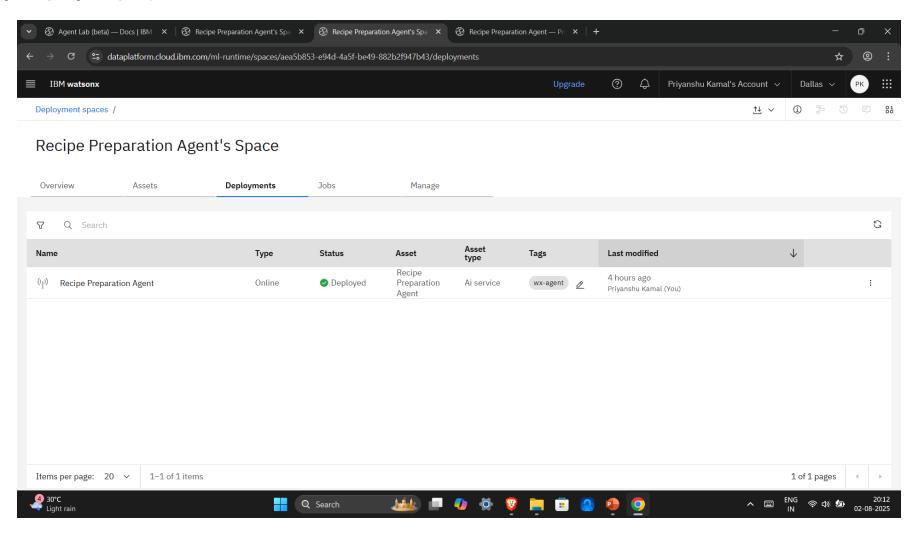
#### Deployment

- The final Al agent is deployed in Watsonx Agent Lab's Deployment Service.
- This allows easy integration into web interfaces or other client applications
- The system supports scalable and real-time interaction



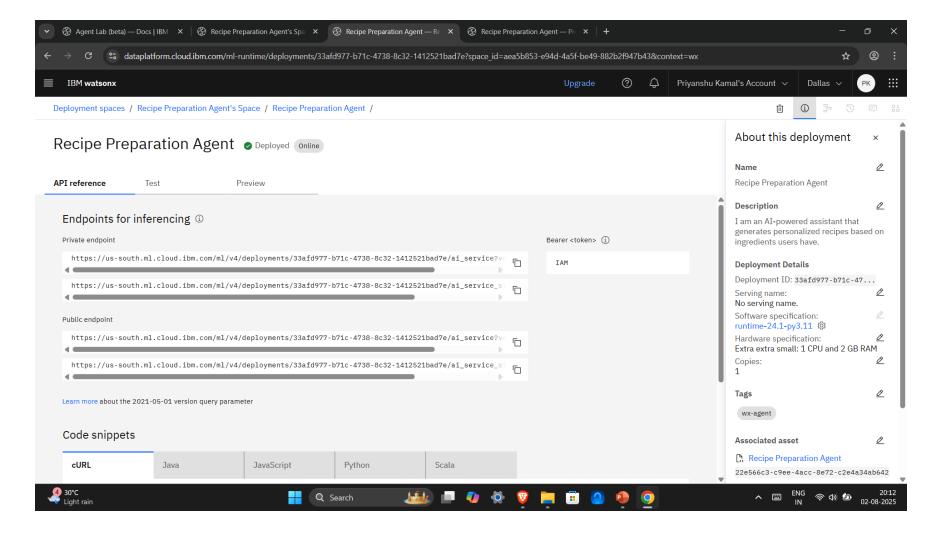
### **DEPLOYMENT PROOF 01/02**

Successfully deployed project.





## **DEPLOYMENT PROOF 02/02**





# RESULT

The Recipe Preparation Agent successfully generates personalized recipe instructions tailored to the user's available ingredients and dietary preferences. The system demonstrates effectiveness in retrieving and adapting recipe content using the RAG approach powered by IBM's Granite foundation model.

#### Observed Outcomes

- The agent retrieves relevant recipes with high semantic match to user-provided ingredients.
- IBM Granite LLM accurately adapts instructions, removes unavailable items, and suggests valid substitutions.
- Users receive complete, step-by-step cooking instructions tailored to their constraints and preferences.

#### Examples of Output

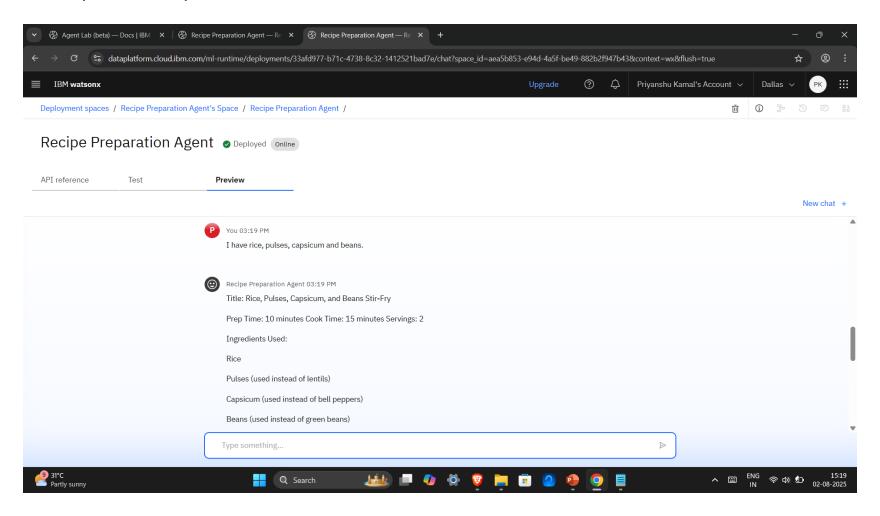
Given: "eggs, tomatoes, onions, rice"
 Output: Tomato Egg Fried Rice with precise steps, suggested substitutions like "use green peas instead of bell peppers", and tips such as "best served with chutney or yogurt."

Visuals in forthcoming slides.



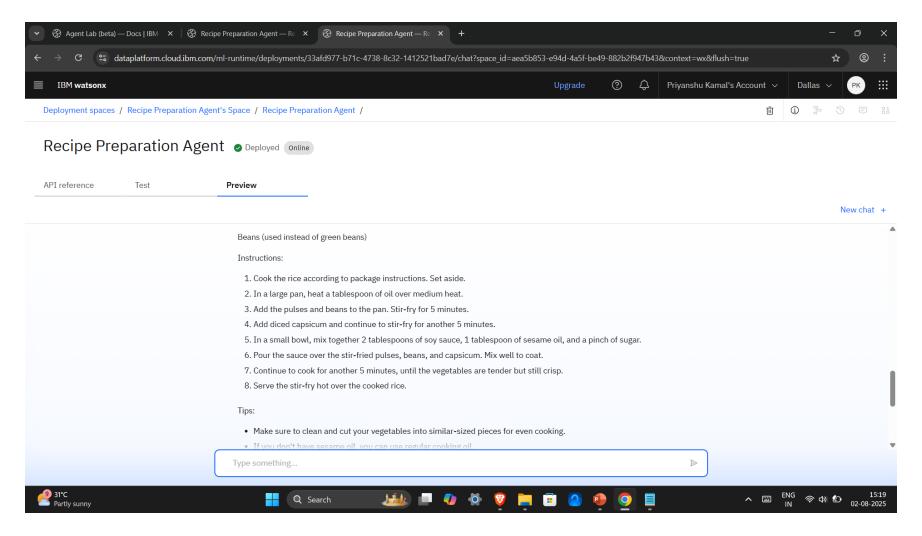
## **TEST RESULT 01 - 1/3**

Input - "I have rice, pulses, capsicum and beans."



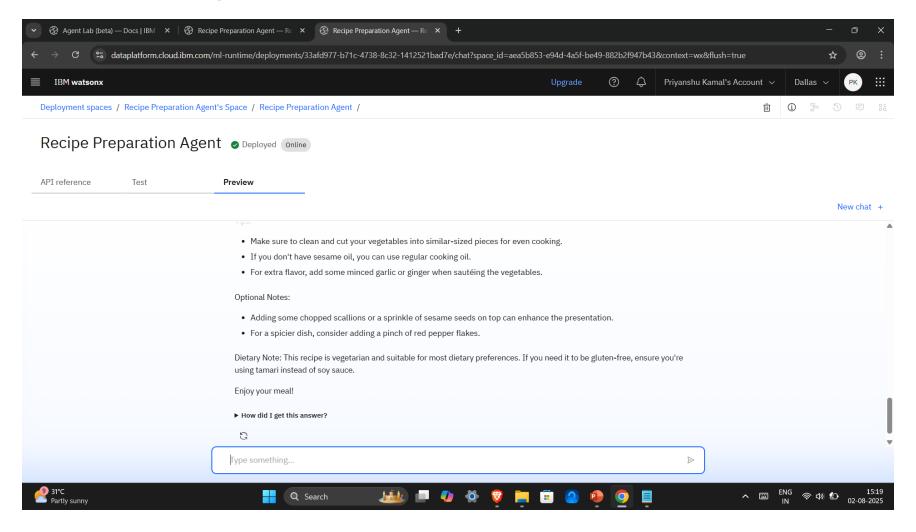


## **TEST RESULT 01 - 2/3**





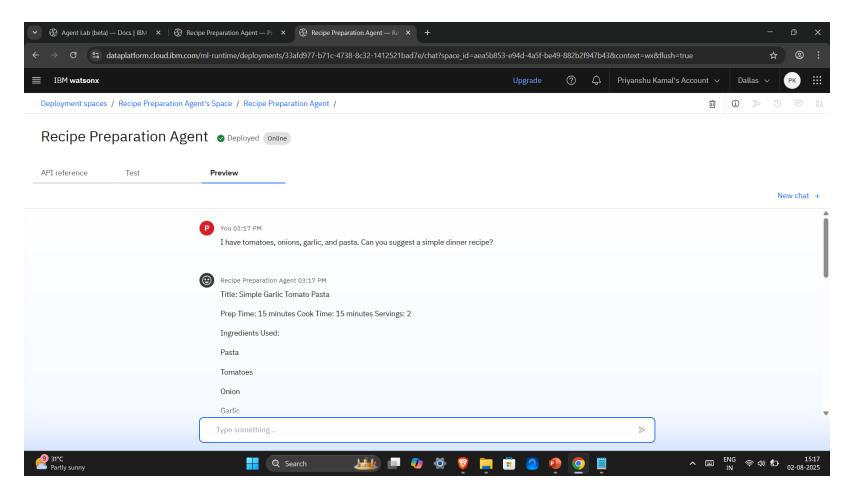
# **TEST RESULT 01 - 3/3**





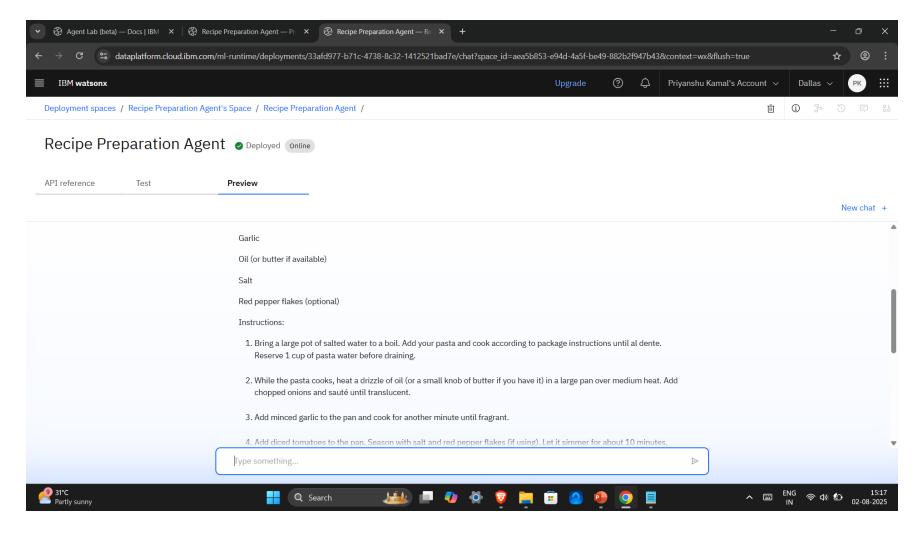
## **TEST RESULT 02 - 1/3**

Input - "I have tomatoes, onions, garlic, and pasta. Can you suggest a simple dinner recipe."



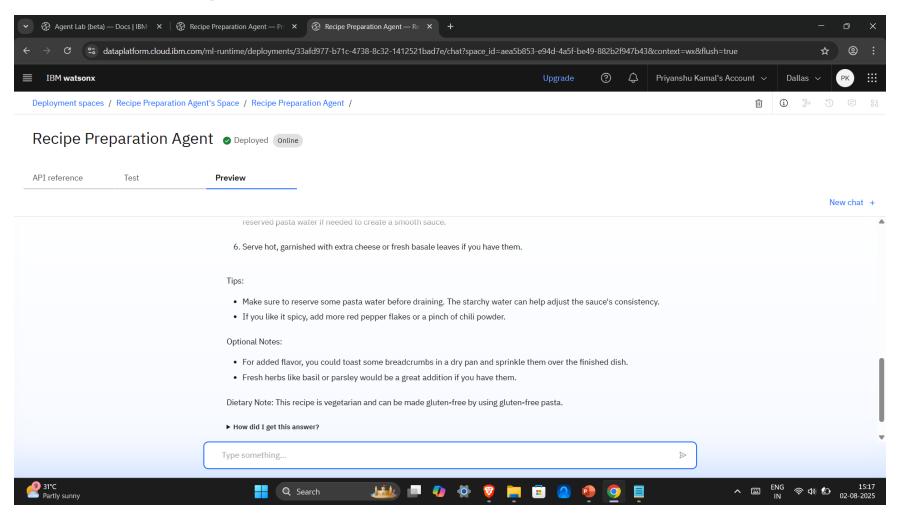


### **TEST RESULT 02 - 2/3**





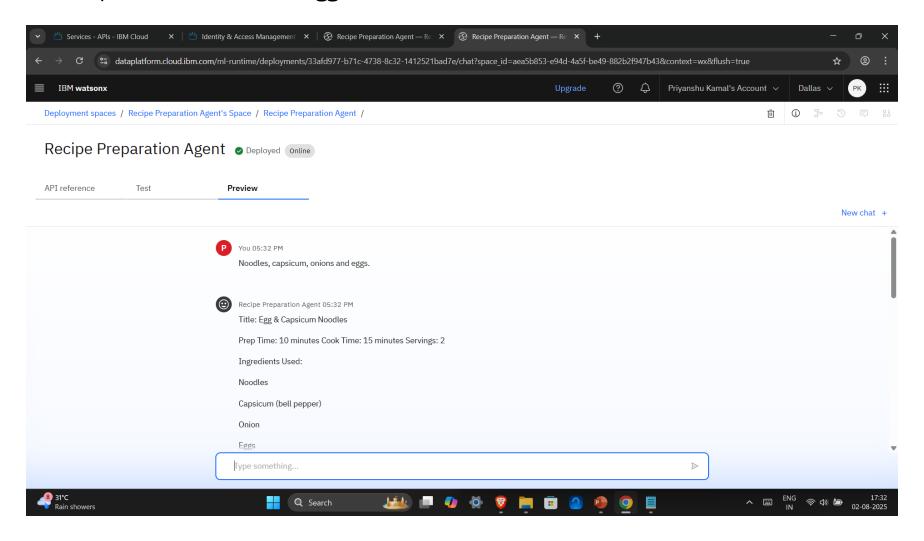
## **TEST RESULT 02 - 3/3**





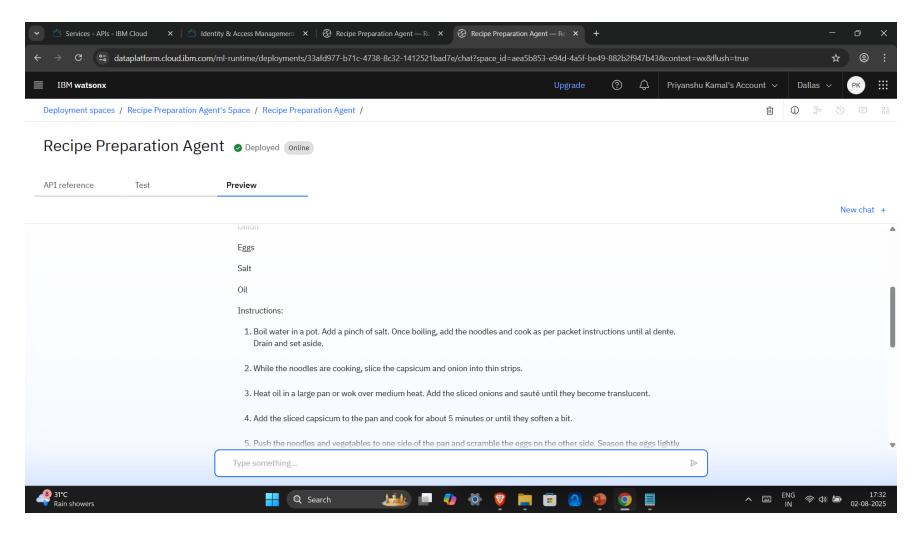
### **TEST RESULT 03 - 1/4**

Input - "Noodles, capsicum, onions and eggs."



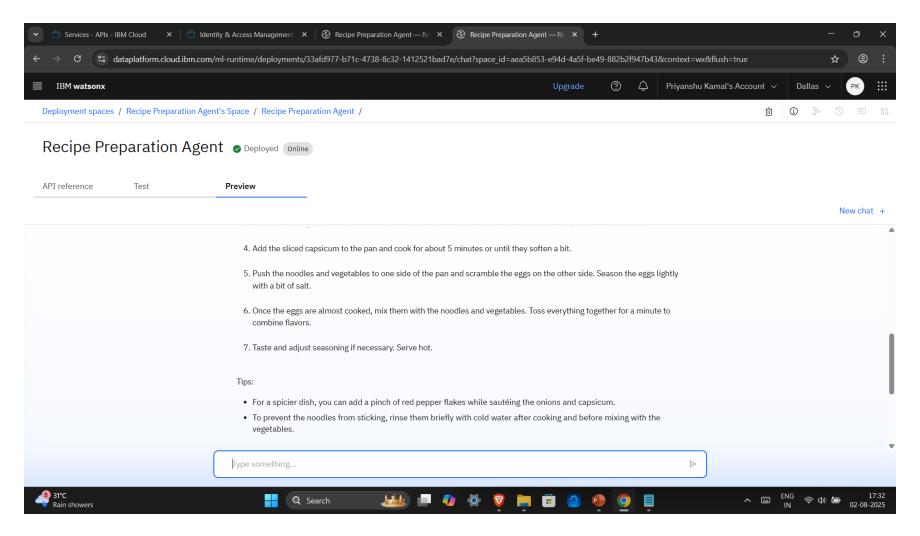


## **TEST RESULT 03 - 2/4**



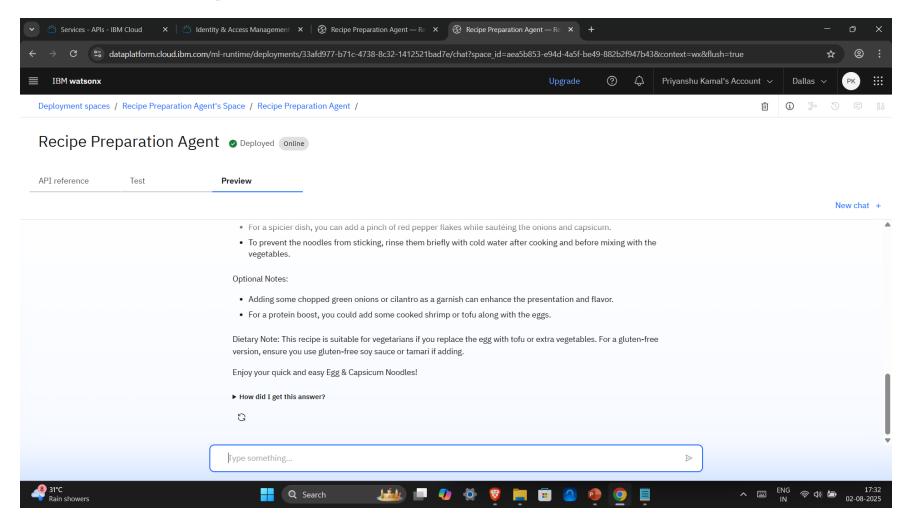


## **TEST RESULT 03 - 3/4**





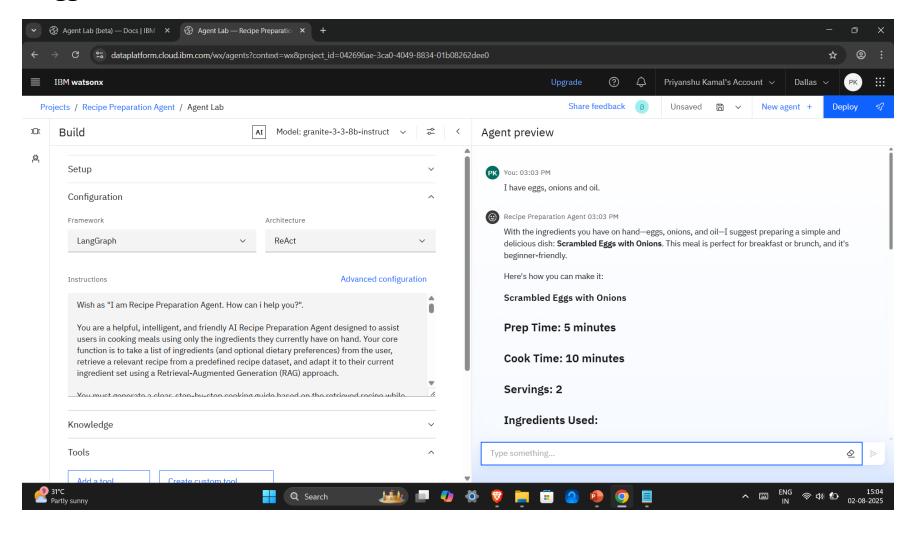
## **TEST RESULT 03 - 4/4**





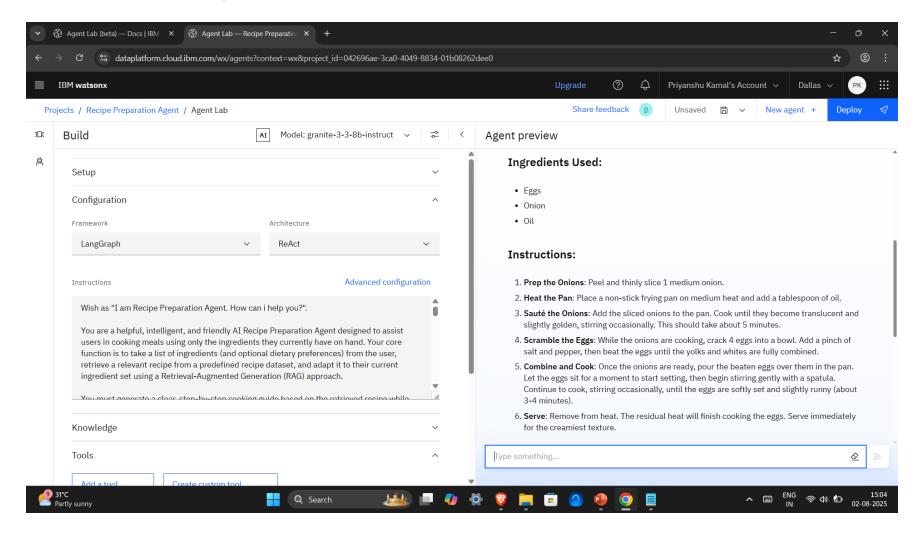
### **TEST RESULT 04 - 1/3**

Input - "I have eggs, onions and oil."



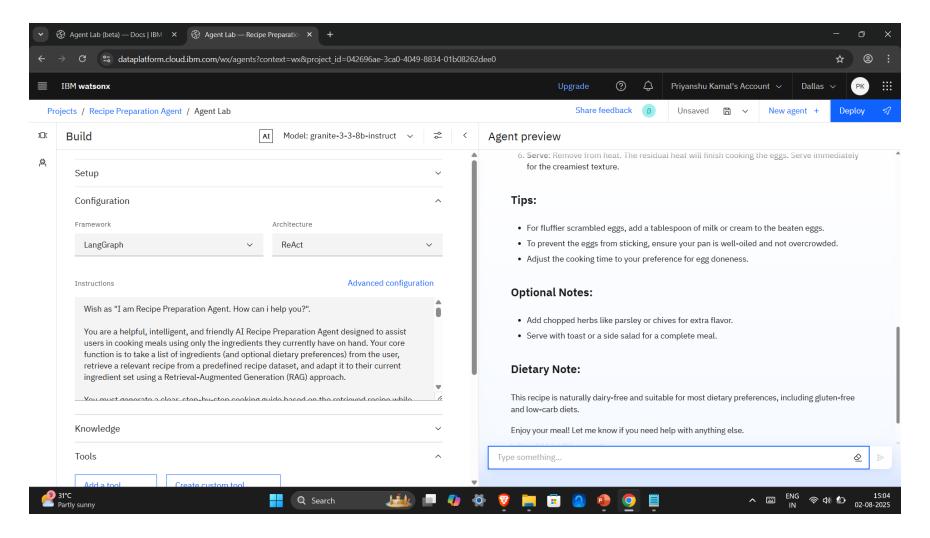


## **TEST RESULT 04 - 2/3**





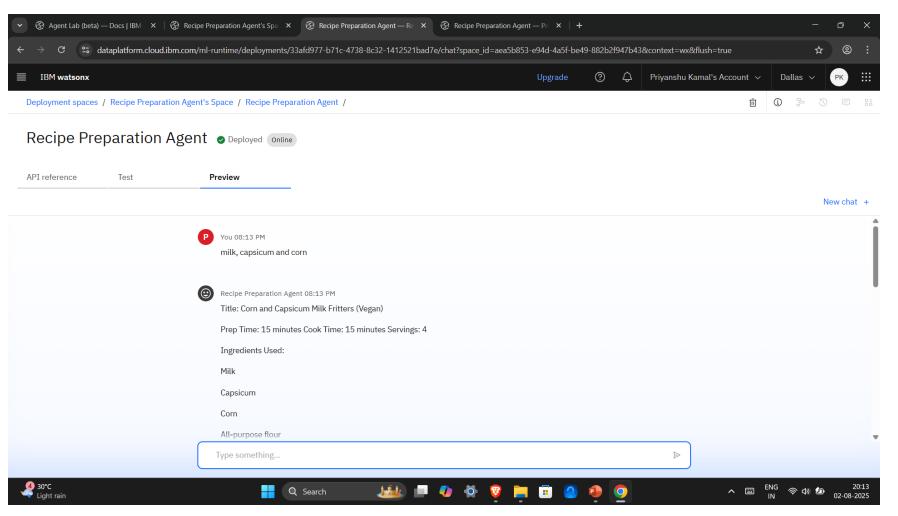
## **TEST RESULT 04 - 3/3**





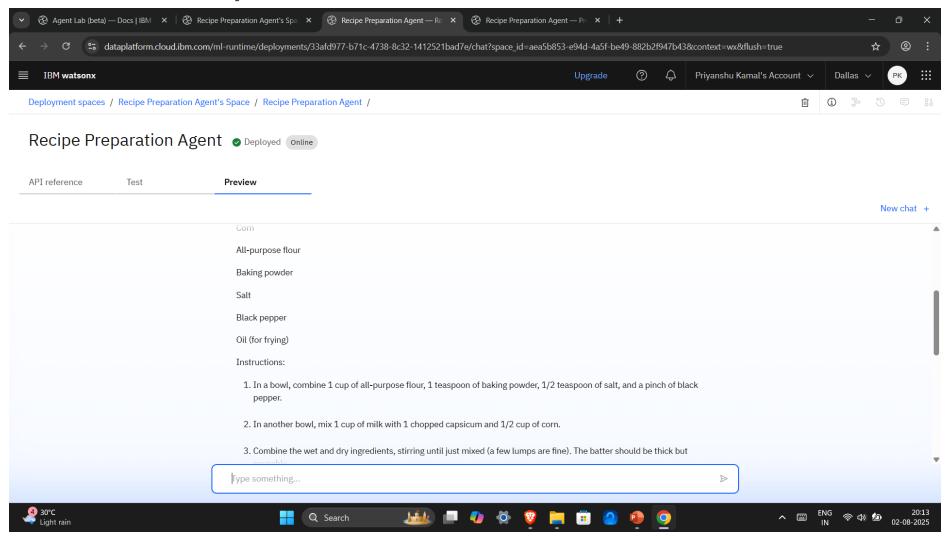
### **TEST RESULT 05 - 1/4**

Input – "Milk, capsicum and corn."



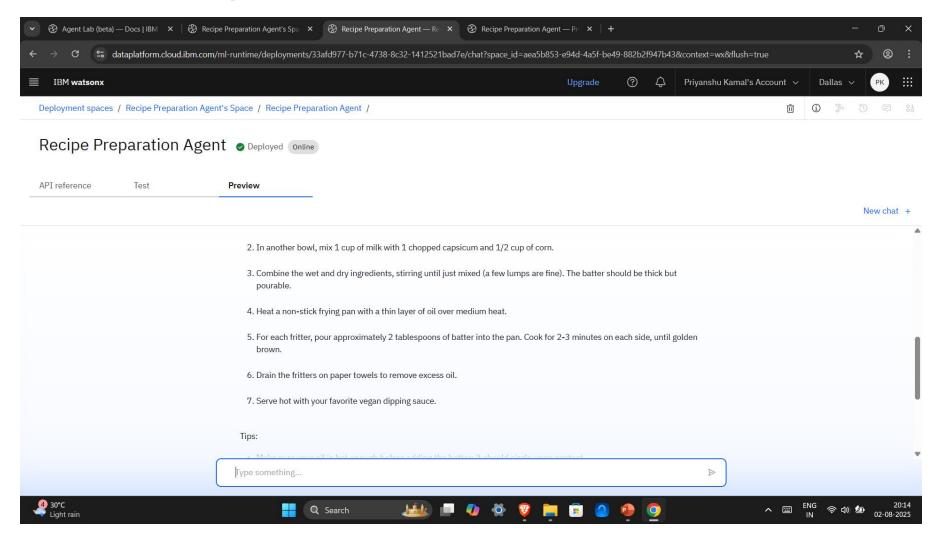


## **TEST RESULT 05 - 2/4**



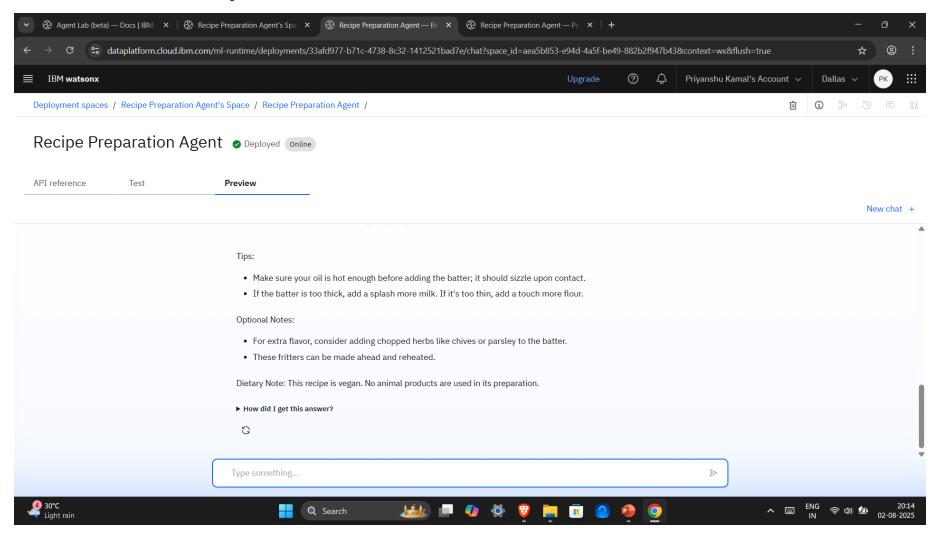


## **TEST RESULT 05 - 3/4**





# **TEST RESULT 05 - 4/4**





# CONCLUSION

- The Recipe Preparation Agent effectively demonstrates how AI can simplify everyday cooking by transforming available ingredients into practical, personalized recipes. By leveraging IBM Cloud's Watsonx, Granite models, and Agent Lab, the system successfully retrieves and adapts recipes in real-time, providing users with step-by-step cooking guidance while considering ingredient limitations and dietary needs.
- This Al-powered approach not only helps reduce food waste and save time but also makes meal preparation more accessible and sustainable.

#### Key Takeaways

- Successfully implemented a RAG-based system using IBM services for intelligent recipe generation.
- Demonstrated that AI can enhance user experience in day-to-day cooking tasks.
- Validated the importance of semantic retrieval and prompt-based adaptation in generating context-aware responses.

#### Challenges & Improvements

- Fine-tuning prompts for highly specific ingredient scenarios
- Expanding recipe diversity to cover more cuisines and dietary categories
- Enhancing evaluation through structured user feedback and analytics



## **FUTURE SCOPE**

The Recipe Preparation Agent holds strong potential for further development and real-world deployment. Future enhancements can significantly expand its
capabilities, improve performance, and enrich the overall user experience.

#### Planned Enhancements

- Dataset Expansion: Incorporate global, regional, and culturally diverse recipes with nutritional data for a broader user base.
- Multilingual Support: Generate and retrieve recipes in multiple languages for international accessibility.
- Voice-Based Interaction: Integrate speech recognition to enable hands-free cooking assistance through voice commands.

#### Technological Advancements

- Real-time Grocery API Integration: Suggest additional ingredients based on local store availability and pricing.
- User Personalization: Build user profiles to provide more personalized recommendations based on history, preferences, and feedback.
- Advanced Retrieval Techniques: Optimize RAG using vector databases and more sophisticated semantic search algorithms.

#### Scalability & Deployment

- Deploy the system as a full-scale web or mobile app.
- Consider integration with smart kitchen devices or IoT platforms.



# REFERENCES

- IBM Watsonx Documentation
   https://www.ibm.com/cloud/watsonx
   Official documentation for Watsonx.ai, Agent Lab, and Granite models.
- IBM Cloud Object Storage
   https://www.ibm.com/cloud/object-storage
   Used to store and access the recipe dataset for retrieval.
- Joseph R. Martinez Recipe Dataset
   GitHub Repository: <a href="https://github.com/josephrmartinez/recipe-dataset">https://github.com/josephrmartinez/recipe-dataset</a>
   Created a smaller dataset out of structured dataset of ~13,000 recipes used as the knowledge base.
- Prompt Engineering Best Practices OpenAl <a href="https://platform.openai.com/docs/guides/prompt-engineering">https://platform.openai.com/docs/guides/prompt-engineering</a>
   <a href="https://platform.openai.com/docs/guides/prompt-engineering">https://platform.openai.com/docs/guides/prompt-engineering</a>
   <a href="https://platform.openai.com/docs/guides/prompt-engineering">Used to refine queries and control output from the Al model.</a>
- IBM Cloud Agentic Al Walkthrough Guide (uploaded as part of this project) Step-by-step resource used for building and deploying the Al agent.



#### **IBM CERTIFICATIONS**

"Getting Started with AI"

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This certificate is presented to

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# Lab: Retrieval Augmented Generation with LangChain

(ALM-COURSE\_3824998)

According to the Adobe Learning Manager system of record

Completion date: 24 Jul 2025 (GMT)

Learning hours: 20 mins

edunet

"RAG Lab"

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# **THANK YOU**

