CAPSTONE PROJECT

RECIPE PREPARATION AGENT USING IBM GRANITE

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OUTLINE

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PROBLEM STATEMENT

• Many users struggle with daily meal planning and waste ingredients they already have. The problem is to suggest personalized recipes based on available ingredients, including dietary needs and substitutions, using Al. The system must retrieve and generate step-by-step cooking instructions, helping reduce food waste and cooking stress.



PROPOSED SOLUTION

- The proposed system is a Recipe Preparation Agent implemented using IBM Cloud services, specifically leveraging the capabilities of IBM Watsonx.ai, Granite foundation models, and LangGraph. The agent follows a Retrieval-Augmented Generation (RAG) architecture to provide personalized cooking assistance based on user-input ingredients.
- The solution architecture consists of the following components:
- **IBM Granite LLM (granite-3-3-8b-instruct):** Used for generating context-aware, step-by-step recipe instructions, substitutions, and dietary recommendations.
- IBM Granite Embedding Model: Converts recipe documents and user input into vector embeddings for semantic similarity search.
- LangGraph with ReAct pattern: Enables a modular and interactive agent framework, allowing the agent to retrieve knowledge and reason over multiple steps before producing a final output.
- Vector Store (FAISS): Stores embedded recipe documents to support semantic search and retrieval based on ingredient matching.
- IBM Watsonx Agent Lab: Used to design, test, and deploy the agent using no-code and low-code tools within IBM Cloud Lite.
- Deployment: The final agent is deployed as a web-based chatbot through IBM Cloud, accessible via a secure public endpoint. It integrates with Watsonx runtime for inference and API orchestration.

SYSTEM APPROACH

- The system is built using IBM Watsonx.ai and follows a Retrieval-Augmented Generation (RAG) architecture. Recipes are embedded using the IBM Granite Embedding Model and stored in a FAISS vector store for efficient retrieval.
- An agent developed using LangGraph with ReAct retrieves relevant recipes based on user input and uses the Granite-3-3-8b-instruct model to generate personalized, step-by-step instructions along with substitutions and dietary adjustments.
- The complete solution is deployed on **IBM Cloud Lite** through **Watsonx Agent Lab**, offering a webbased interface for users to interact with the system.



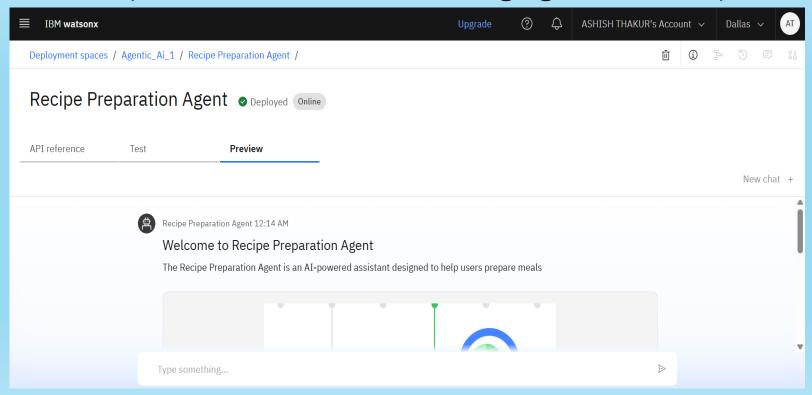
ALGORITHM & DEPLOYMENT

- Algorithm:
- The system uses a Retrieval-Augmented Generation (RAG) pipeline:
- Input Processing: User provides available ingredients and preferences.
- Embedding: Input and recipe dataset are embedded using the IBM Granite Embedding Model.
- Retrieval: Relevant recipes are fetched via FAISS based on semantic similarity.
- Generation: Retrieved data and user context are passed to Granite-3-3-8b-instruct to generate:
 - Step-by-step recipe instructions
 - Ingredient substitutions
 - Cooking tips and dietary adjustments
- Deployment:
- The agent is created and tested in Watsonx Agent Lab using LangGraph (ReAct pattern).
- Deployed as a web-based Al assistant on IBM Cloud Lite.
- Provides a publicly accessible interface for users to interact with the agent in real-time.

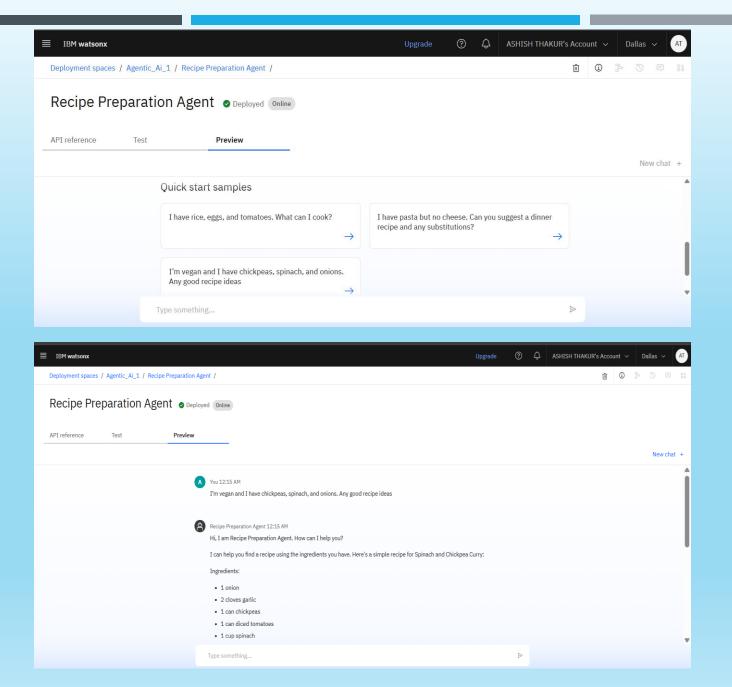


RESULT

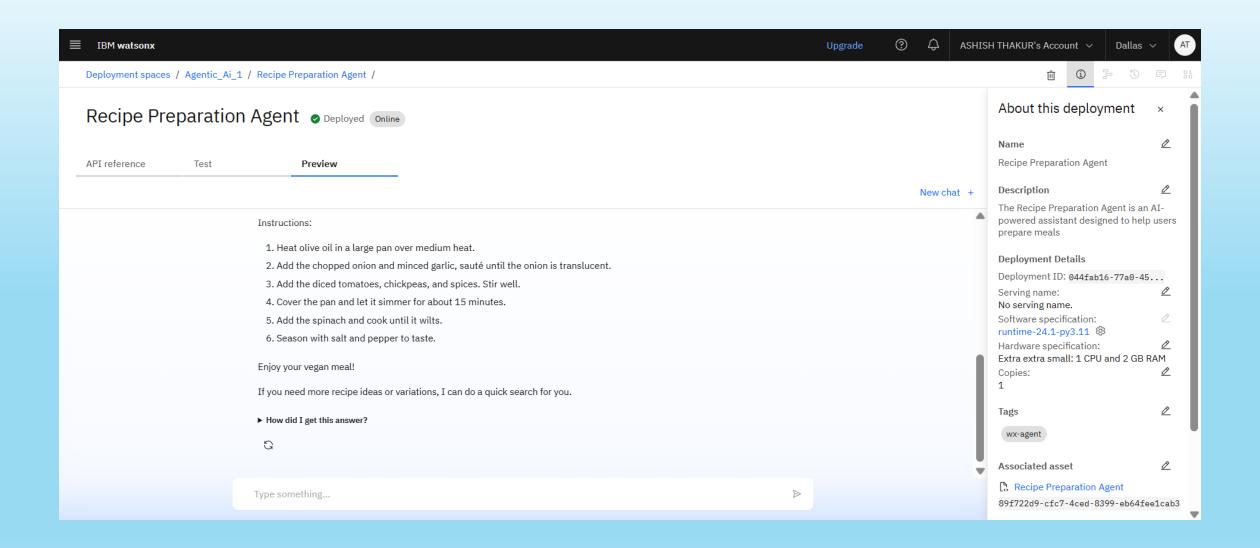
The system successfully returns personalized recipes based on user-provided ingredients. It suggests creative substitutions and considers dietary preferences (e.g., vegan, gluten-free). Screenshot outputs show step-by-step instructions and interactive chat interface via Watsonx. Include visualizations and comparisons between predicted and actual counts to highlight the model's performance.













CONCLUSION

The Recipe Preparation Agent demonstrates how AI can simplify daily cooking, reduce food waste, and provide useful dietary support. Using IBM Granite and LangGraph, the agent delivers meaningful recipe guidance in real time. The system is scalable and adaptable for integration into smart kitchen apps. Highlight any challenges encountered during the implementation and potential improvements. Emphasize the importance of accurate bike count predictions for ensuring a stable supply of rental bikes in urban areas.



FUTURE SCOPE

- Future enhancements include:
- Integrating voice input for hands-free cooking support
- Expanding the recipe database with multilingual and regional recipes
- Real-time inventory syncing with kitchen IoT devices
- Personal health tracking with nutrition-aware recipe adjustments
- Offline support for edge devices



REFERENCES

- •IBM Watsonx Documentation https://www.ibm.com/watsonx
- •IBM Granite Models IBM Research Foundation Models
- LangGraph & LangChain https://docs.langchain.com
- •FAISS: Facebook AI Similarity Search https://github.com/facebookresearch/faiss
- •Recipe Dataset (optional source) Kaggle: Food.com Recipes and Reviews Dataset
- •Retrieval-Augmented Generation (RAG) Facebook AI Research Blog



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Completion Certificate



This certificate is presented to

Ashish Thakur

for the completion of

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



THANK YOU

