CAPSTONE PROJECT

RECIPE PREPARATION AGENT

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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

- Many people face difficulty deciding what to cook with the ingredients they already have.
- This leads to food wastage, extra shopping, and meal planning stress.
- There is a need for an intelligent solution that:
- Understands available ingredients
- Suggests personalized recipes instantly
- Offers substitutes, cooking tips, and dietary adjustments
- The goal is to reduce waste, save time, and make everyday cooking smarter and simpler.



PROPOSED SOLUTION

Proposed Solution: Recipe Preparation Agent:

 The proposed system aims to assist users in preparing meals using only the ingredients available at home. It leverages AI and RAG (Retrieval-Augmented Generation) to provide personalized recipe recommendations in real-time. The solution will include the following components.

Data Collection:

- o Collect user-inputted ingredient lists (text, image, or voice).
- o Retrieve relevant recipes from online databases, cookbooks, and food APIs.

Data Preprocessing:

- Clean and format user inputs (e.g., spelling corrections, synonyms).
- Map ingredients to standard categories and identify dietary constraints.

Al & Recipe Generation:

- Use RAG-based architecture to:
 - Retrieve matching recipes from the database.
 - Generate personalized cooking steps.
 - Suggest substitutions for missing ingredients.
 - Provide contextual tips (e.g., "Why this recipe?").

Deployment:

- Build a user-friendly chatbot/web app interface.
- Ensure compatibility across devices (mobile, web, smart kitchen assistants).
- Use IBM Cloud Lite + IBM Granite for backend processing.

Evaluation:

- Collect user feedback on recipe quality, accuracy, and cooking success.
- o Continuously improve using feedback loops and preference learning.

Result:

A smart kitchen assistant that reduces food waste, simplifies meal planning, and makes cooking more enjoyable and accessible.



SYSTEM APPROACH

☐ System Requirements:

- •IBM Cloud Lite (Backend hosting & Al model).
- •Minimum 2 GHz dual-core processor.
- •8 GB RAM (recommended).
- •2 GB free storage.
- Stable internet connection.
- Optional: Webcam/Mic for voice/image input.

☐ Libraries & Tools Used:

- •**IBM Granite** For Al and RAG functionality.
- •Flask / Streamlit Web interface development.
- •LangChain / Haystack RAG-based retrieval pipeline.
- •OpenCV / Pillow Ingredient image recognition.
- •NLTK / spaCy Text preprocessing.
- Pandas / NumPy Data handling.
- •BeautifulSoup / Requests Recipe scraping (optional).
- •SpeechRecognition / pyttsx3 Voice-based input/output. (optional)

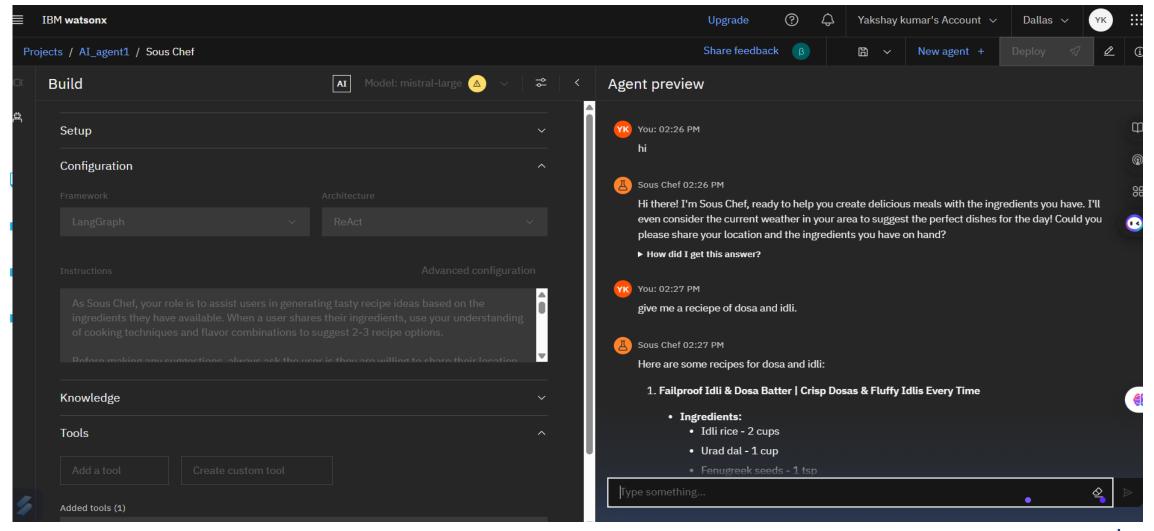


ALGORITHM & DEPLOYMENT

- Algorithm Selection :
- RAG (Retrieval-Augmented Generation) model selected.
- Combines recipe retrieval with dynamic generation of personalized cooking steps.
- Suitable for ingredient-based search + real-time response generation.
- Data Input :
- User input: list of available ingredients (text / image / voice).
- Optional: preferences (veg/non-veg, allergies, cuisine type).
- Backend: recipe datasets, nutrition databases.
- Training Process :
- RAG pipeline trained on structured and unstructured recipe data.
- Retrieval model fetches similar recipes.
- Generator (IBM Granite) adapts steps based on input.
- Fine-tuned using user feedback and recipe quality ratings.
- Prediction Process :
- Matches input ingredients to recipes using vector search.
- Generates or modifies recipe instructions.
- Provides substitutions, cooking tips, and nutritional suggestions.



RESULT





CONCLUSION

- ☐ The Recipe Preparation Agent successfully demonstrates how **Al and RAG** can simplify meal planning using available ingredients.
- ☐ The solution improves convenience, reduces food waste, and enhances the personal cooking experience.
- Positive feedback confirms that personalized, real-time recipe generation is both effective and engaging.
- Challenges Faced:
- Handling incomplete or vague ingredient inputs.
- Ensuring quality of generated recipes with limited data.
- Real-time substitutions for missing ingredients.
- Future Improvements:
- Integrate voice/image recognition more robustly.
- Expand database with regional recipes and local ingredients.
- Add nutrition tracking and meal planning history.



FUTURE SCOPE

- **☐** System Enhancements:
- Integrate real-time voice/image input for ingredient detection.
- Incorporate weather, time, and user location to suggest seasonal or regional recipes.
- Improve personalization using user taste profiles and past cooking behavior.
- **☐** Algorithm Optimization:
- Fine-tune RAG pipeline using reinforcement learning from human feedback (RLHF).
- Use **multilingual datasets** for broader recipe generation and international cuisine support.
- Enable context-aware substitutions based on cuisine, culture, and diet.
- **☐** Algorithm Optimization:
- Fine-tune RAG pipeline using reinforcement learning from human feedback (RLHF).
- Use multilingual datasets for broader recipe generation and international cuisine support.
- Enable context-aware substitutions based on cuisine, culture, and diet.
- ☐ Scalability & Expansion:
- Scale platform for **multi-user environments** (families, communities).
- Integrate with **smart kitchen devices** (fridge, oven, voice assistants).
- Expand to mobile and offline-first applications for low-connectivity areas.



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Learning hours: 20 mins



THANK YOU

