

HACKEREARTH RAG APPLICATION – SPRINT 2

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Introduction

We will develop a RAG (Retrieval-Augmented Generation) application for HackerEarth that will utilize vector search, knowledge graphs, and a LLM to answer questions and generate content from a knowledge base of more than 10,000 Wikipedia articles.

Sprint Objectives

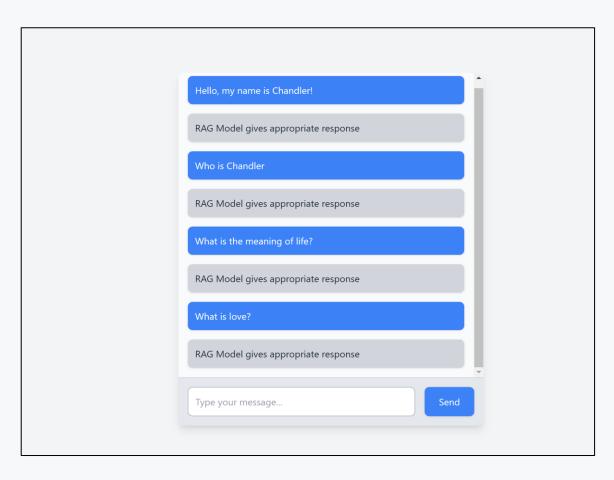
- Embeddings Creation and Vector Search Preparation: To implement efficient text embeddings, we researched and tested embedding methods, specifically focusing on BERT-based embeddings for use in the vector search module.
- Frontend Prototype Development: Developing a minimalistic chatstyle interface to enable seamless interaction between users and the RAG application.
- Knowledge Graph Querying: Initial setup and testing of DBpedia integration to enable semantic data retrieval for more accurate responses.
- System Architecture Refinement: Documentation updates on the system architecture to ensure a clear breakdown of subsystems, including NLP, embeddings, and data processing workflows.

Feature Implementation

Team Member	Feature	Objective
Ethan	Embeddings	Embeddings Creation and Vector Search Preperation
Chandler	Frontend	Frontend Prototype Development
Molly	Knowledge Graph	Knowledge Graph Querying
Adam	Natural Language Processing	Query Processing

Figma Prototype & Client Feedback

Minimalistic
 chat/messaging system
 with the RAG model



Kanban Overview & Contributions

Summary:

Molly

- Knowledge Graph code
 Architecture diagram and subcomponent decomposition report sections
 Unit and Integration testing report sections
 Project Milestone document

Ethan

- Text embeddings research and code
 Architecture introduction and system overview report sections
- Testing strategy and environment requirements report sections
- Sprint report

Chandler

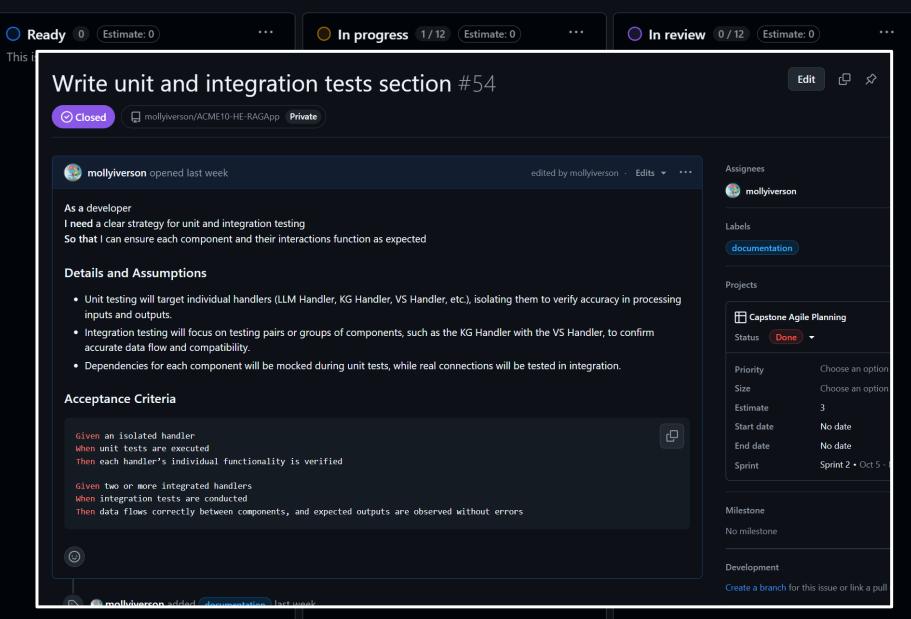
- React Frontend prototype
 User Interface design report section
- System testing report section (functional, performance, and user acceptance testing)
 Python code for extracting and embedding Wikipedia
- data

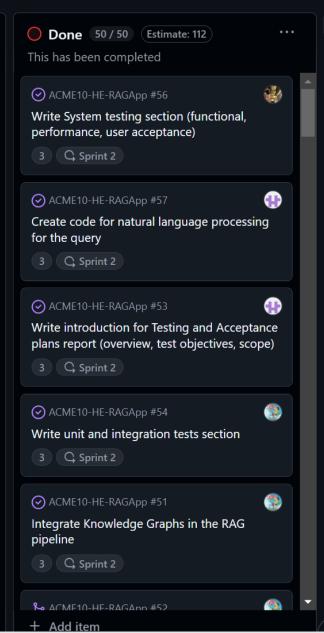
Adam

- Natural Language query processing code
 Data Design and some subcomponent report sections
 Testing overview, test objectives, and scope report
- sections
- Python notebook with FAISS tutorial

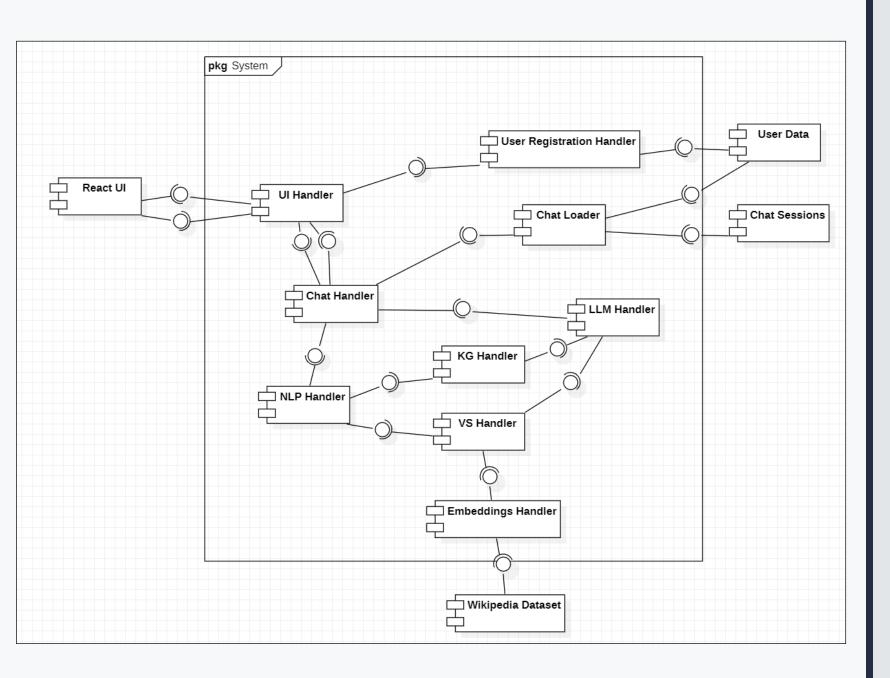
All

Meeting notes for client meetings





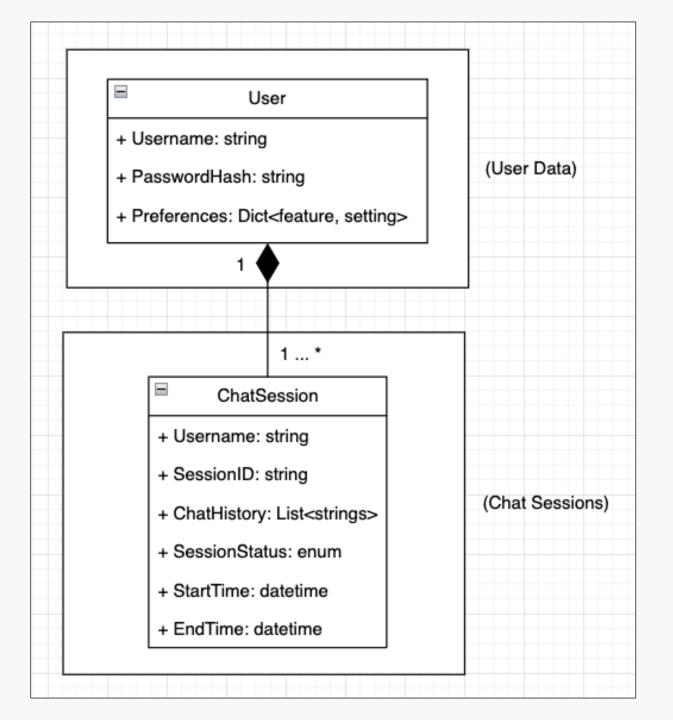
Documentation of System Architecture, Database, UI, & Testing



System Architecture

Technologies chosen:

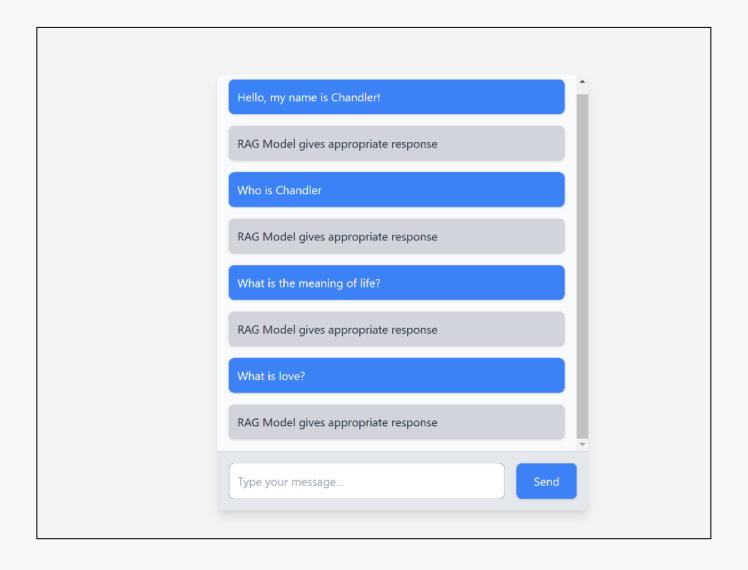
- React for UI
- BERT for embedding
- spaCy and NLTK for NLP
- DBpedia for KG
- FAISS for VS
- LLaMA for LLM



Database Design

Optimization strategies:

 Indexing on Username and SessionID



User Interface Design

Unit

- Testing isolated handlers
- Covering edge cases and boundary conditions

Integration

- Grouping interconnected components to test data flow
- Mocking outside components and dependencies

System

- Manual functional testing
- Measuring response time
- Stress testing
- Gathering feedback from stakeholders and end users

Testing Plan

Tools:

- Unittest or PyTest as a testing framework
- unittest.mock for mocking dependencies
- Coverage.py to measure code coverage
- spaCy and NLTK to compare query responses
- GitHub Actions for CI/CD
- Jmeter or Locust to assess response time and scalability

Sprint Achievements & Challenges

Looking back on Sprint 2.

- Solution Approach Section
- Testing Approach Section

- Extracting Wikipedia dataset
- NLP processing of a query
- Generating text embeddings for sample of dataset
- DBpedia querying
- Basic frontend prototype of RAG App

Next Steps & Retrospectives

- Upcoming Tasks:
- Sprint Retrospective
- Client Feedback Consideration
- Generate text embeddings
- Integrate vector search
- Integrate knowledge graph
- LLM processing for responses

Conclusion

- Sprint 2 Achievements:
- Next Steps:
- Optimize Embedding Generation: Scale up for the full Wikipedia dataset with GPU acceleration.
- Implement FAISS Indexing: Enable fast vector search for relevant results.
- Link to Knowledge Graph: Enhance response accuracy by integrating vector search with the knowledge graph.
- o **Refine LLM Processing**: Improve query responses by combining vector and knowledge graph data.
- Advance Frontend: Continue developing the chat interface for a smoother user experience.

Demo