

HACKEREARTH RAG APPLICATION – SPRINT 3

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Introduction

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

Pipeline Integration (Main Objective): Establish a fully functional query-processing pipeline that integrates vector search (VS), large language model (LLM), natural language processing (NLP), and knowledge graph (KG) components.

Vector Search and LLM: Develop a system that retrieves contextualized results using vector search and processes them through an LLM.

Integration of Components: Combine outputs from the knowledge graph and vector search with the LLM in the query pipeline.

NLP and KG Integration: Automate SPARQL query generation from NLP output and connect the NLP and KG modules to the system interface.

CI/CD Pipeline: Implement a CI/CD pipeline to automate testing development workflows.

Feature Implementation

Team Member	Feature	Objective
Ethan	Vector SearchComponentLLM Component	 To get contextualized queries with VS results Run queries through an LLM and receive a clear response
Chandler	-Vector Search and LLM Integration	 Combine KG and VS outputs to form a contextualized query Run the frontend with the LLM integrated
Molly	- NLP and KG Integration	 Take NLP output to create a SPARQL query for KG Connect the NLP and KG pipeline to the frontend
Adam	- CI/CD Pipeline for Testing Automation	- Automate development testing

Feature Demos

Testing and Validation

Integration Tests

- NLP tasks
 - Tokenization
 - Entity extraction from a query (for KG querying)
 - Harmful intent check

Unit Tests

- Knowledge graph
 - DBPedia querying
- Vector search
 - Loading embeddings
 - Building vector index
 - Search

Kanban Overview & Contributions

Molly

- In-Class Presentation
- Integration of NLP and KG pipeline
- Future work report section

Ethan

- Client Presentation
- Vector Search and LLM code
- Alpha prototype description report section

Chandler

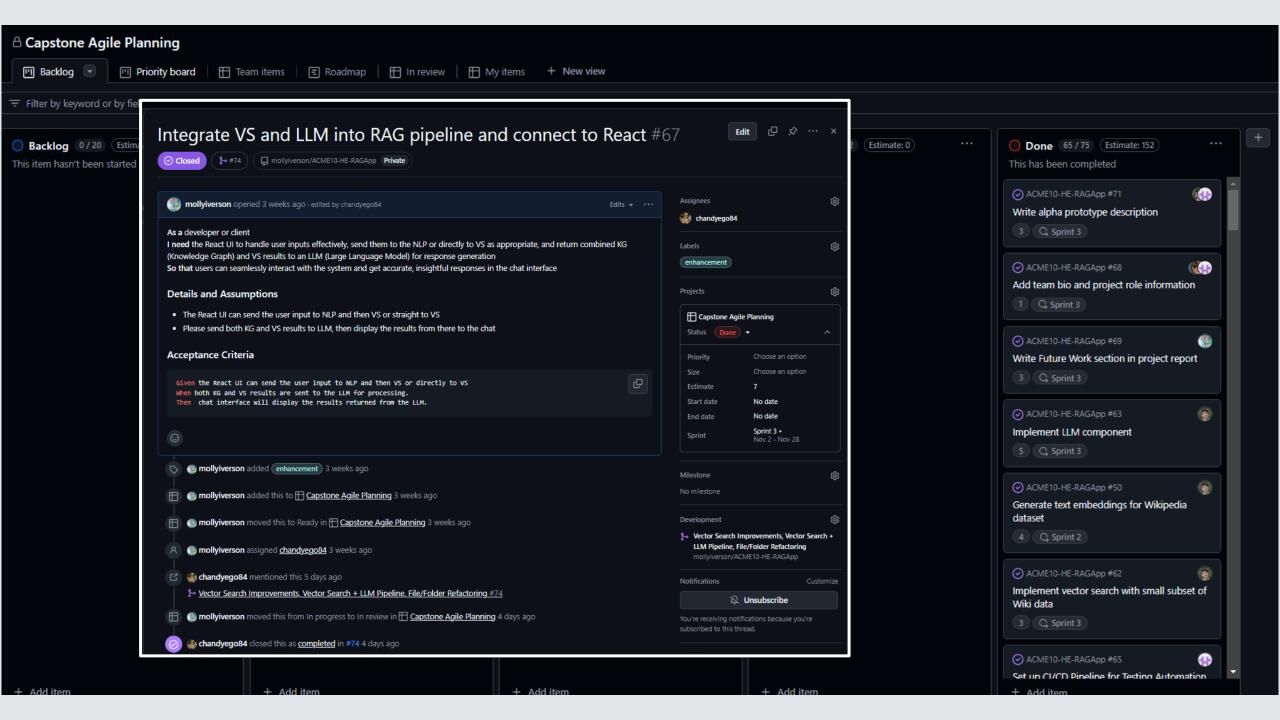
- In-Class Presentation
- Integration of VS and LLM pipeline
- Alpha prototype demo report section

Adam

- Client Presentation
- CI/CD pipeline for testing automation
- Alpha prototype description report section
- Sprint report

All

Meeting notes for client meetings



Documentation of Prototype Report

Sprint Achievements & Challenges

Achievements

- Alpha Prototype Description Completed:
- Finalized and condensed the description within the report's page constraints.
- Moved excessive details to appendices to maintain clarity and focus.
- Implemented LLM Handler:
- Leveraged LLaMA for generating natural language responses.
- Successfully combined Knowledge Graph (KG) and Vector Search (VS) outputs.
- Vector Search Functionality Integrated:
- Used FAISS to optimize search results based on Wikipedia embeddings.
- CI/CD Pipeline Setup:
- Automated dependency installation and testing via GitHub Actions.
- Improved testing consistency across the development process.
- Backend and Frontend Integration:
- Linked VS, LLM, and KG components to the React UI.

Sprint Achievements & Challenges

Challenges and Solutions

- Challenge: Integration of VS and LLM into the RAG pipeline.
- Solution: Modularized handlers and streamlined the data flow to ensure compatibility.
- Challenge: Optimizing query responses from the LLM for context relevance.
- Solution: Implemented cosine similarity and debugging messages for better accuracy.

Sprint Achievements & Challenges

Progress from Sprint 2

- Enhanced vector search with cosine similarity for improved accuracy.
- Established an LLM endpoint and integrated KG,
 VS, and React UI for a cohesive pipeline.

Sprint Retro

What Went Well:

- Initial prototype demonstrated core RAG functionality
- Robust testing: unit, integration, and automated testing with CI/CD pipeline

Improvements Needed:

See Client Feedback

Client Feedback



Feedback:

- Focus on RAG pipeline, not fullstack features
- Ensure scalability for 10,000+ embeddings
- Enhance response speed and context relevance

Actions Taken:

- Removed plans for user authentication and chat session storage
- Adjusted next semester's plans

Next Steps

Upcoming Tasks:

- Generate and Store 10,000 embeddings
- Implement chunking for better data embedding and retrieval
- Test the system with custom input
- System testing
- Explore different approaches to the retrieval part of RAG
- Improve NLP

Conclusion

Key Progress:

- Alpha prototype successfully demonstrated core capabilities
- Plans adjusted to prioritized system optimization and new retrieval approaches

Future Direction:

- Embed full Wikipedia dataset
- Optimize RAG pipeline performance

Thank You!

Please let us know if you have more questions or suggestions