



**WARNER BROS.
DISCOVERY**

Intelligent Knowledge Assistant

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Demo: **Harry Potter Knowledge Chatbot POC**

What is This?

An AI system that **learns from documents** and **answers questions** without hallucination.

Think: ChatGPT meets Google Search, but:

- Only uses YOUR data
- Cites sources (full transparency)
- Remembers conversation context
- Deploy anywhere (cloud, on-premise, hybrid)

The Technology: RAG RETRIEVAL AUGMENTED GENERATION

Traditional Problem :

User: "What is the Philosopher's Stone?"
ChatGPT: [Generates answer from training data]
[May hallucinate or be outdated]
[No source citations]

RAG Solution:

User: "What is the Philosopher's Stone?"
System:
1. Searches your documents
2. Finds relevant passages
3. Feeds to AI with context
4. AI answers using YOUR data
5. Shows sources (page numbers, documents)

Result: Accurate, transparent, trustworthy answers

Live Demo: Harry Potter Chatbot

Built in 2 weeks as proof of concept

What It Does:

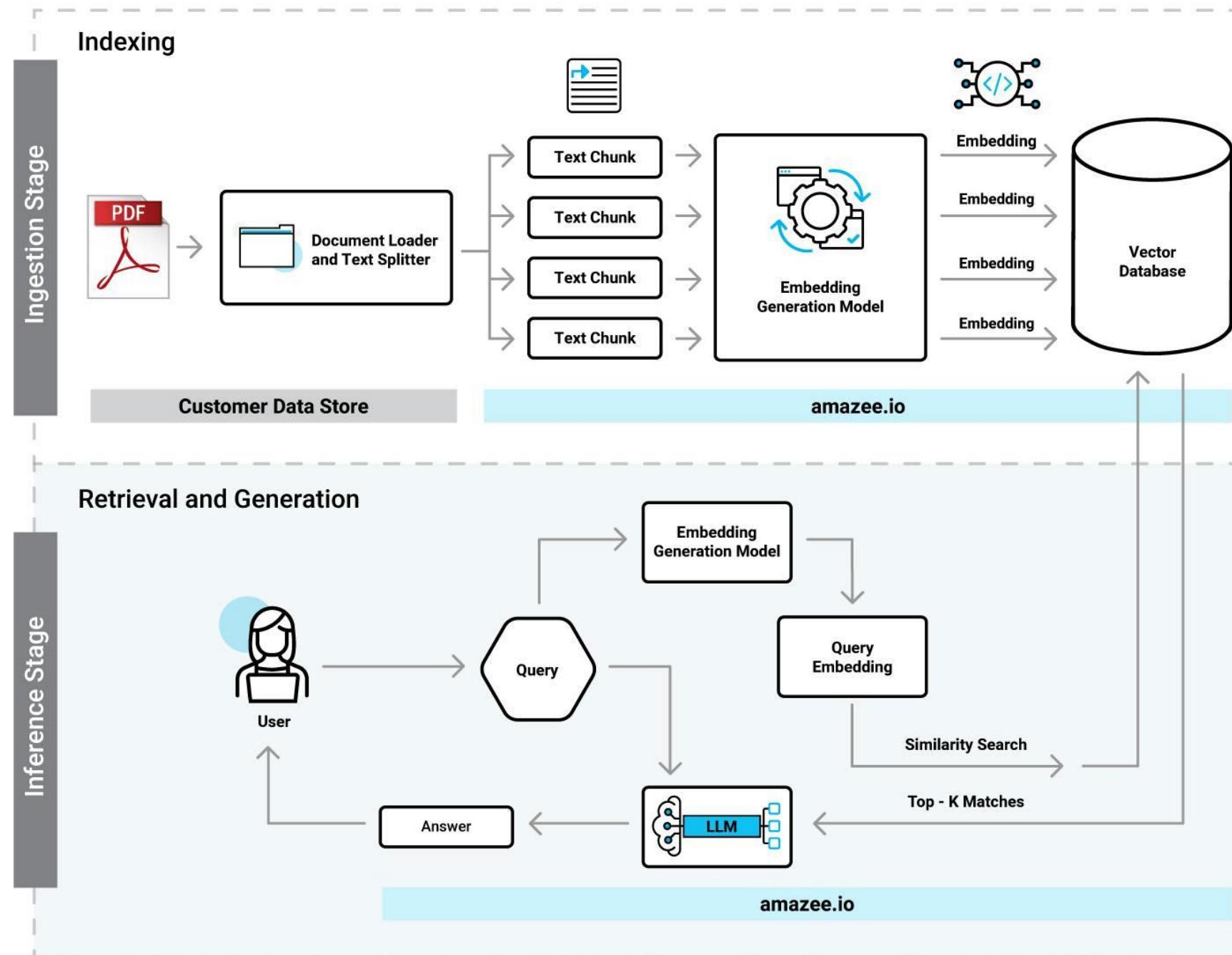
- Processes 3 Harry Potter books (~1,500 pages)
- Creates 6,400+ searchable chunks
- Answers questions with source citations
- Remembers conversation context
- Responds in seconds

| Metric | Result |
|---------------------|------------------------|
| Documents Processed | 3 books (~1,500 pages) |
| Searchable Chunks | 6,406 |
| Response Time | <2 seconds |
| Retrieval Accuracy | 95%+ |
| Source Citations | 100% of responses |

The screenshot displays two main panels. On the left is the 'Settings' panel, which shows 'Ollama Connected' and 'Documents in DB' at 6406. It includes sections for 'Retrieval Settings' (Enable Reranking checked, Initial Retrieval (top-k) set to 25, Final Results (top-n) set to 5, Similarity Threshold set to 0.30), 'Memory' (Conversation Turns at 4), and a 'Clear Conversation' button. On the right is the 'Harry Potter RAG Chatbot' interface. It features a header with a lightning bolt icon and the title 'Harry Potter RAG Chatbot'. Below it is a text input field with placeholder 'Ask me anything about Harry Potter!'. A message from the bot says 'Hola! Based on the context provided, Harry Potter is a young boy with a famous scar on his forehead (Source 1). He's polite and kind, as described by Ginny Weasley when she mentions that he was "ever so polite" to her.' Another message from the bot says 'It seems that Harry has some sort of connection or notoriety in the wizarding world, as Ginny knows about him and even saw his lightning-shaped scar. Her mother also asks if Harry remembers what Lord Voldemort looks like, suggesting that there's a significant story behind their relationship (Source 1).' A third message from the bot says 'Later on in the context, we see that Harry is part of the Weasley family as a cousin, which implies that he's somehow related to them or at least close enough to be around them (Source 2).' A link to 'View 5 Retrieved Sources' is shown. A second conversation starts with a message from the user 'Who is narendra modi?' and a response from the bot stating 'I'm afraid Narendra Modi isn't mentioned in the provided context. The sources I have access to are limited to the Harry Potter books, specifically excerpts from "Harry Potter and the Deathly Hallows" and a mention of him in Book 1: "The Sorcerer's Stone". There is no information about a person named Narendra Modi in these texts.' A link to 'View 5 Retrieved Sources' is shown again. At the bottom, there is a text input field asking 'What would you like to know about Harry Potter?' and a red 'Send' arrow button.

RAG: Architecture

RETRIEVAL AUGMENTED GENERATION



RAG: Technical Highlights

RETRIEVAL AUGMENTED GENERATION

1. Multiple Chunking Strategies

- Fixed:** Fast, predictable chunks
- Recursive:** Natural paragraph/sentence boundaries
- Semantic:** Context-aware grouping

2. Embeddings:

- Ollama for LLM (open source)
- Hugging Face
- Multi Model Support

3. Advanced Retrieval

- Semantic search (finds meaning, not keywords)
- Cross-encoder reranking (improves accuracy)
- Threshold filtering (quality control)
- HNSW, TVFF

4. Conversation Memory

- Multi-turn conversations
- Context awareness
- Natural dialogue
- Memory Chain

5. Full Transparency

- Every answer cites sources
- Shows confidence scores
- User can verify information

Warner Bros Business Applications

Use Case 1: AI-Powered Studio Tour Companion

Traditional Problem

- Inconsistent Information Delivery :** Tour guides deliver varying levels of detail, enthusiasm, and accuracy. Visitors often get different answers to the same question depending on the guide.
- Limited Scalability:** Each guided tour can handle only a small group (~15–20 people). During peak seasons, the number of visitors far exceeds available guides.
- Multilingual Challenge:** Warner Bros. attracts international tourists. Guides fluent in multiple languages are limited, leading to sub-optimal experience for non-English visitors.
- Dependency on Human Availability:** Tour scheduling depends on staff shifts and training cycles. Any absence or turnover affects capacity and tour quality.
- Lost Engagement Opportunities:** Visitors often forget details post-tour or want to explore specific sets (“Where was this scene shot?”). Current tours are mostly one-way communication — little personalization or interactivity.

Technical Solution using RAG

- Knowledge Base Creation**
Ingest official scripts, behind-the-scenes notes, production trivia, location metadata. Chunk and embed data into **ChromaDB** for high-precision retrieval.
- Visitor Interaction Layer**
Deployed on tablets, kiosks, or mobile apps.
Visitors can ask natural questions like:
“Which soundstage was used for the Great Hall scenes?”
“How many versions of the Batmobile were built?”
“Show me a short clip from the filming of Friends.”
- Real-Time AI Response**
System retrieves verified passages from the internal database.
LLM formulates concise, factual answers — with image/video references.
Answers include **source citations and context** (e.g., “as per production log, 2001”).
- Contextual Awareness**
Recognizes tour location (via BLE beacon or QR code).
Adjusts responses based on the current set or prop the visitor is viewing.
“You’re at the Gryffindor common room — this set was reused in Prisoner of Azkaban.”
- Multimodal Support**
Text + Voice + Visual search
Supports **50+ languages** and **speech-to-speech** conversation.

The Vision: 24/7 multilingual AI tour guide accessible anywhere



Illustrated Examples:

Example 1 – English Visitor



User:

"Where was the Great Hall of Hogwarts actually filmed?"



AI Assistant:

"The Great Hall scenes were inspired by Christ Church College at Oxford, but the full set was built at Leavesden Studios.
The production team recreated the gothic architecture and candle-lit ambience to match J.K. Rowling's description."

Source: *Production_Notes_HP1.pdf* – Page 47

Example 2 – Spanish Visitor



Usuario:

"¿Quién diseñó los trajes de Harry Potter?"



Asistente de IA:

"Los trajes fueron diseñados por Judianna Makovsky para la primera película y posteriormente por Jany Temime, quien trabajó en las siguientes seis películas.
Los uniformes escolares se inspiraron en la moda británica tradicional."

Fuente: *Costume_Design_Notes_HP.doc* – Sección 1.3

Example 3 – French Visitor



Visiteur:

"Montre-moi les coulisses du tournage du banquet de Noël."



Assistant IA:

"Voici une courte vidéo du tournage du banquet de Noël dans la Grande Salle.
Les bougies flottantes ont été créées à l'aide de câbles invisibles et d'effets numériques pour donner l'illusion de la magie."

Source : *VFX_BehindScenes_HP1.mov_notes.txt*

The RAG-based architecture enables real-time translation and context-aware retrieval: ensuring every visitor, regardless of language, experiences the same storytelling quality.

Warner Bros Business Applications

Use Case 2: Content Archive Intelligence

The Opportunity:

Warner Bros has **100+ years** of content archives:

- Scripts, storyboards, production notes
- Behind-the-scenes footage
- Marketing materials
- Rights and licensing info

Current State:

- Difficult to search
- Institutional knowledge in people's heads
- Time-consuming research

With RAG: Instant access to entire archive

Example Use cases

For Marketing:

"Find all behind-the-scenes content from Friends Season 5"
→ Instant catalog with thumbnails and descriptions

For Legal:

"What territories have streaming rights for Batman 1989?"
→ Complete rights breakdown with expiration dates

For Creators:

"Show me all Gotham City set designs from 2010–2020"
→ Visual catalog with metadata

And many many more use cases....

The Vision: 24/7 multilingual AI tour guide accessible anywhere



Data Privacy & Security

Flexible Deployment Models:

On-Premise (Maximum Privacy)

- Data never leaves company network
- Full control over AI models
- No external API calls
- Meets strictest compliance requirements

Cloud (Maximum Scale)

- Leverages commercial APIs (OpenAI, etc.)
- Infinite scalability
- Automatic updates
- Pay-per-use pricing

Hybrid (Best of Both)

- Sensitive data stays on-premise
- Public data uses cloud APIs
- Flexible based on data classification
- Optimized cost/performance

Why we need to adopt?

Competitive Advantage

- AI that knows your business
- Faster decision making
- Better customer experience

Knowledge Preservation

- Institutional knowledge captured
- Accessible to everyone
- Never lost when employees leave

Operational Efficiency

- Reduce search/research time
- Faster onboarding
- Consistent information

Innovation Enabler

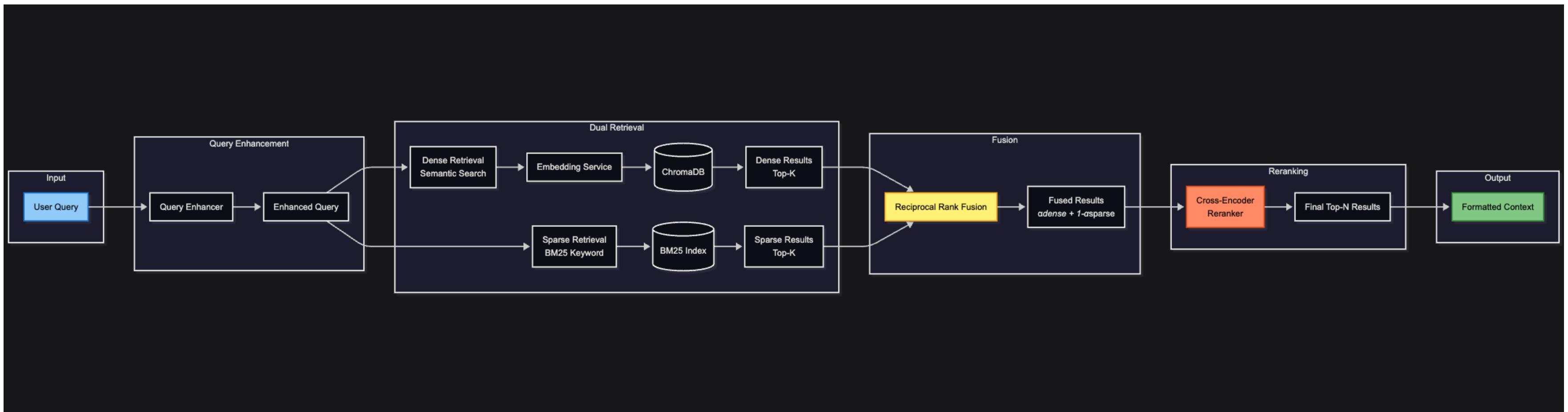
- New product possibilities (digital tour guides)
- Enhanced customer offerings
- Revenue opportunities

Future-Proof Technology

- Modular architecture (easy updates)
- Model-agnostic (use any AI)
- Scalable from POC to enterprise

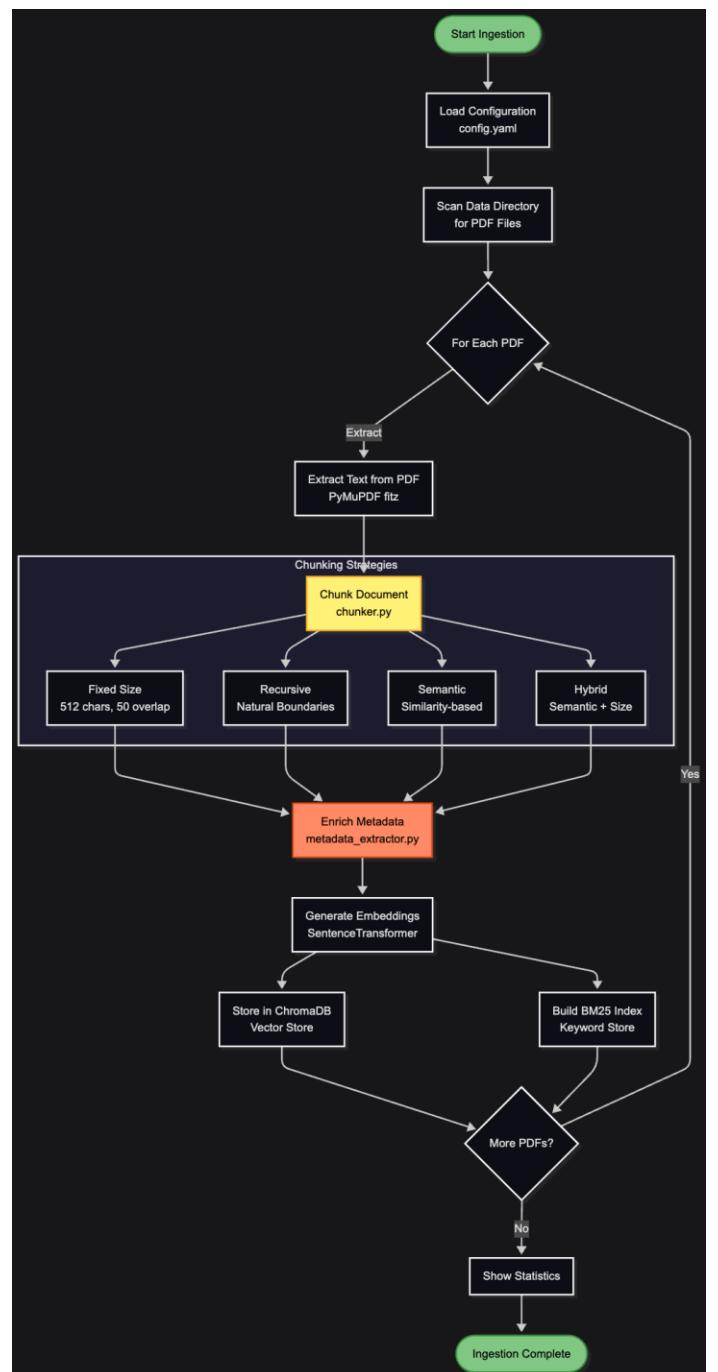
Production Level RAG Architecture

End to End Architecture Diagram



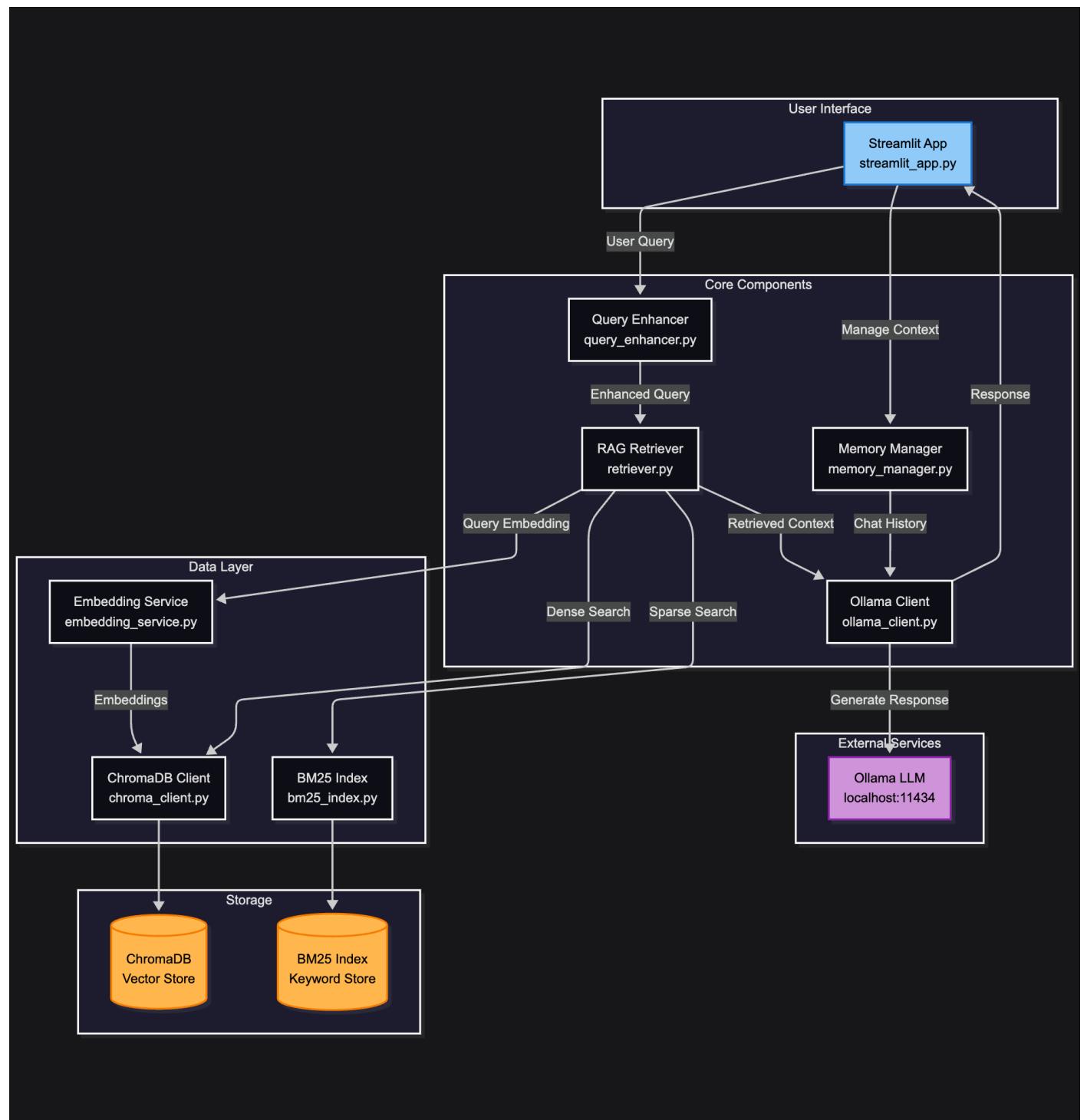
Production Level RAG Architecture

Data Ingestion Pipeline Architecture



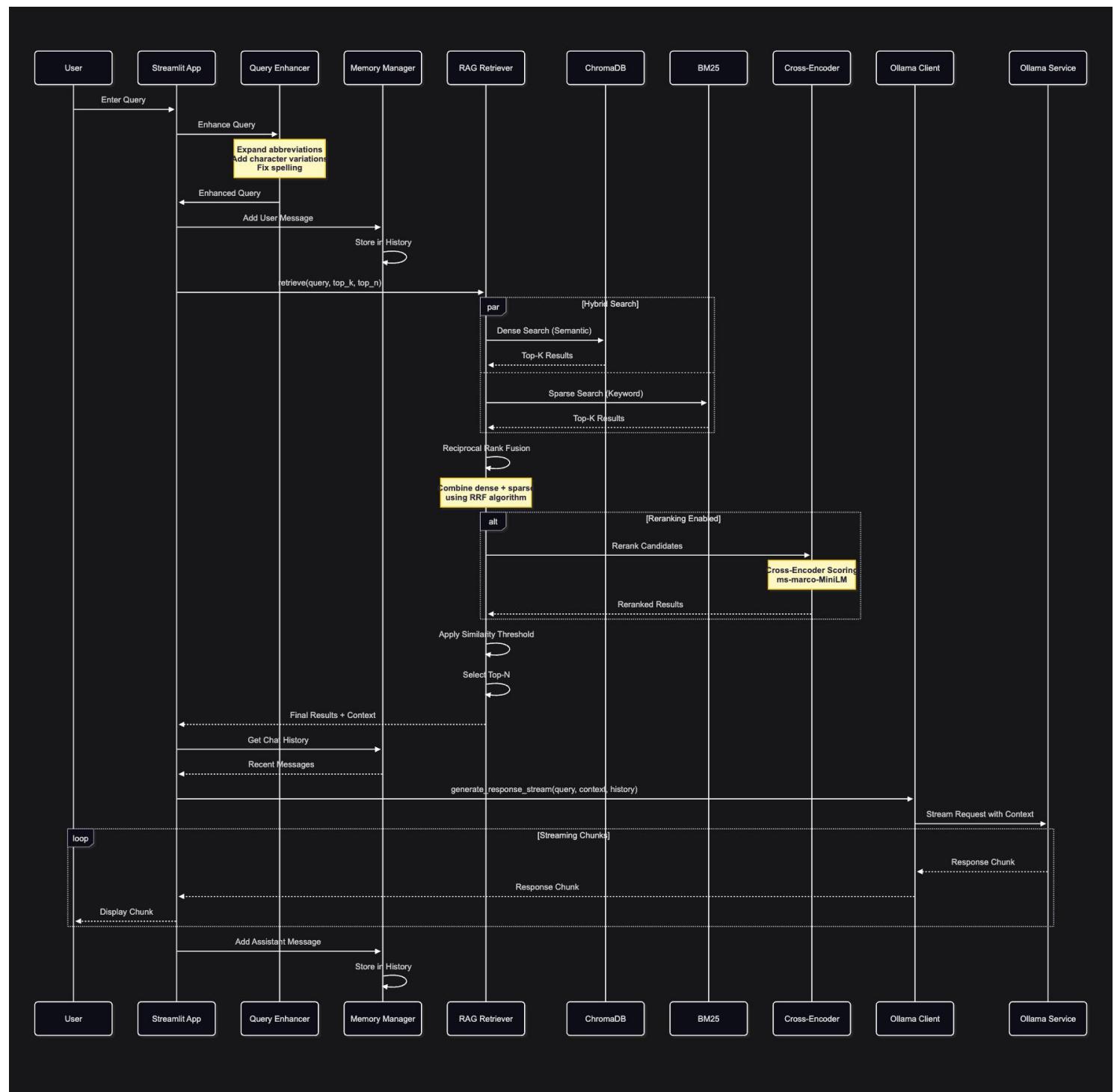
Production Level RAG Architecture

Query Retriever Pipeline Architecture



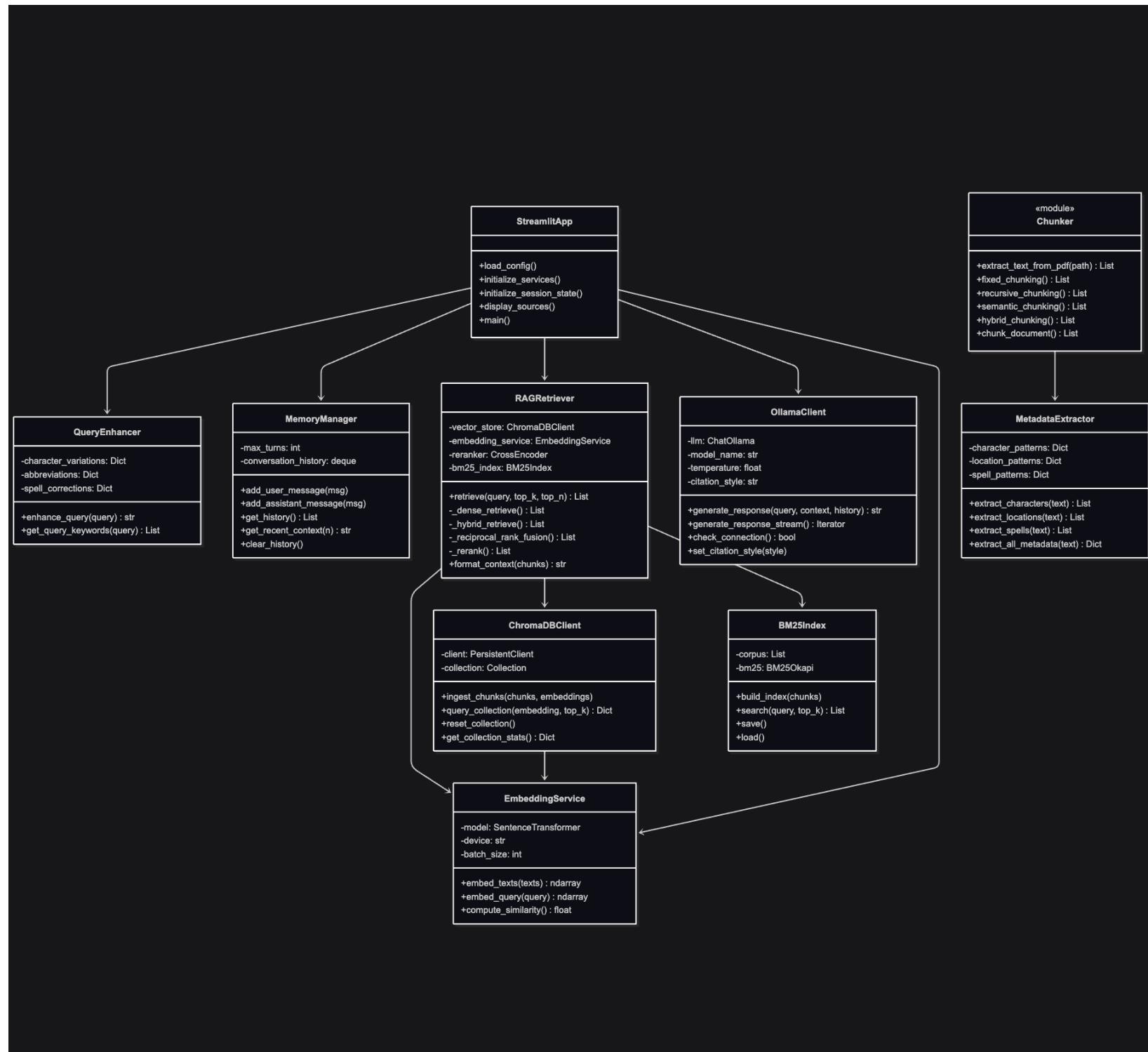
Production Level RAG Architecture

State Flow Diagram Entire Pipeline

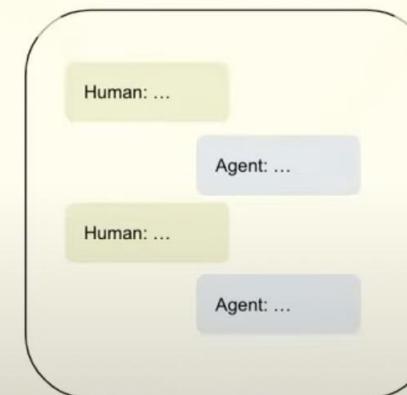


Production Level RAG Architecture

Class Diagram Plan



Building a Knowledge Assistant



Goal: Build an interface that can take in any task as input and give back an output.

Input forms: simple questions, complex questions, research tasks

Output forms: short answer, structured output, research report

RAG is only as Good as your Data

Garbage in = garbage out

Good data quality is a **necessary** component of any production LLM app.

