GRCResponder

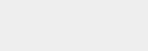
Brianna Steier, Liam Gass, Elijah Tavares, Cael Howard, June Kim, Rish Sharma, Angel Li

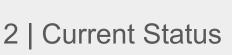


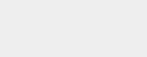


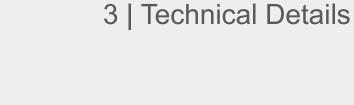
Table of Contents

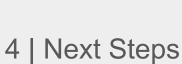
1 | Overview













Problem Statement

- What is a General Rate Case proceeding?
 - Used to address costs of operation and maintenance for utility companies
 - o Carried out by the California Public Utilities Commission
- Utility companies need to be able to
 - Manage documents
 - Respond to regulatory inquiries
 - Ensure consistent responses across submissions
 - Meet strict deadlines and compliance requirements



Overview

GRC Responder

- ChatBot that assists with legal questions related to general rate case proceedings
- Provides historical examples and supporting documentation to quicken the search process
- Summarizes key points from lengthy documents

Retrieval-Augmented Generation

- Given a particular query, be able to find relevant context based on a vector similarity search across all stored databases
- o Pass retrieved context to LLM model in order to form a response to return to the user



Current Status

Web Scraping

- Scraped all proceeding metadata from Utility Commission website
- Capable of individual proceeding document processing and database insertion

To-Do

 Optimizing document collection efficiency while balancing scraper politeness

```
| Debunk_metadata | Debunk_text | Debunk_metadata | Debunk_metadat
```





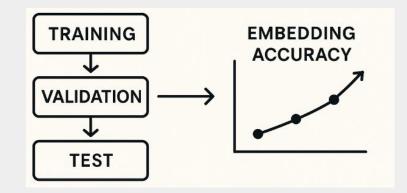
Development Updates

Threading

- To improve efficiency we are experimenting with batch processing and threading
 - Batch size of 3 document cases
 - Operating over 4 threads

Other Focuses

- Creating a validation set for embedding model and query accuracy.
 - Batch size of 3 document cases
 - Operating over 4 threads





Current Status

LLM Integration

- Connected LLM to both local and azure-hosted chroma database
- Performed tests using earlier, available version of web scraper

Evaluation

 Set up initial testing system using more powerful LLM to judge responses

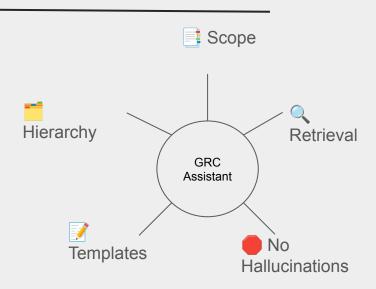
```
Human Message ==========
What documents do you have access to?
  ----- Ai Message =-----
 retrieve (60d8d6c8-b2eb-4123-a015-395ab48132ef)
Call ID: 60d8d6c8-b2eb-4123-a015-395ab48132ef
   query: all available documents
Retrieved 2 documents in 2.6690289974212646 seconds.
                 ========= Tool Message =========
Name: retrieve
Source: {'chunk_index': 0, 'document_id': 4, 'document_type': 'APPLICATION', 'filed_by': 'San Diego Gas & Electr
Content: Document: Application of San Diego Gas & Electric Company (U902M) for Authority to Establish Its Author
Type: APPLICATION
Filed By: San Diego Gas & Electric Company
Filing Date: 2025-03-20
Source: {'chunk_index': 0, 'document_id': 3, 'document_type': 'AMENDMENT', 'filed_by': 'San Diego Gas & Electric
Content: Document: Amendment to Test Year 2026 Cost of Capital Application of SDG&E (U902M) dated March 20, 202
Filed By: San Diego Gas & Electric Company
Filing Date: 2025-03-24
                   ------ Ai Message ------
I have access to two documents:
1. An "Application" document with the title "Application of San Diego Gas & Electric Company (U902M) for Authori
2. An "Amendment" document with the title "Amendment to Test Year 2026 Cost of Capital Application of SDG&E (U9)
Please note that these documents are related to a specific proceeding, with proceeding numbers and IDs. If you're
       LLM + Retrieval call took 39.72 seconds
```

Prompting Azure Chroma DB





LLM Updates



System Prompt Updates:

- Scope tightened to CPUC GRC only polite refusal for off-topic requests.
- Explicit three-level evidence order user docs → vector
 DB → static memory; asks follow-ups if data missing.
- Response templates standardized Q&A, consistency table, or draft filing with inline citations.
- Hallucination guardrails added aim to reduce fabricated facts; numerical values cross-checked or flagged "unknown."
- Tone & formatting defined concise, professional, no jargon; should end with "Need more detail?" prompt.



Current Status accenture

Technical Details

Tech Stack:

- TypeScript with React frontend
 - PDFViewer react-pdf-viewer > PDF.js
- Python (FastAPI) backend
- Microsoft Azure

Document Retrieval:

Semantic Search

- Vector Storage: ChromaDB for text embedding storage and retrieval
 - Used to locate relevant documents used to address user queries

Topic Based Navigation

- postgreSQL: document and proceeding metadata storage
 - Uses GIN indexing to map unique words to text chunks
 - Allows for retrieval of only chunk text containing query words





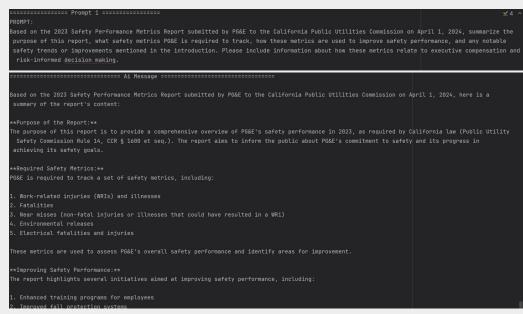
Technical Details

Local Database Setup

- Web scraper targeted entirety of A.21-06-021 and A 22-04-016
- 1000 char chunks w/ 50 char overlap by default

Evaluation

- Using more powerful LLM for evaluation
- For a prompt to local LLM: save prompt, context, and response
- Ask it to evaluate relevance of retrieved context and how well our local LLM extracted information from provided context
- Exploring possibility of generating a small Q&A dataset on relevant documents



Prompting Local Chroma DB





Next Steps...

- Refine RAG Testing & Evaluation:
 - Improve retrieval and context usage by the LLM
 - Evaluation metrics:
 - i. Search response time < 2 seconds
 - ii. Document processing accuracy > 95%
 - iii. >90% accuracy in relevant document retrieval
- Host Project on Azure:
 - Connect web with LLM and have it hosted on Microsoft Azure.
- Enhance User Experience:
 - Polish performance and visuals based on user testing.
- Research Paper:
 - o Document progress and performance metrics in a research paper potentially catered towards a SAGE journal.



Questions?

