# GRCResponder

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





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### ICS Expo Recap

- Presented yesterday at ICS expo
- Pending assignments for the class
  - report and final demo



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#### The Problem

Utility companies regularly undergo General Rate Case (GRC) proceedings with the California Public Utilities Commission (CPUC) to justify the rates they charge their customers. GRC teams must manually sift through thousands of pages of documents to craft responses, which is tedious, repetitive, and inefficient.

#### Our Approach

The GRCResponder is an A.I. chatbot that leverages modern semantic search technology to revolutionize this process. Using vector database embeddings, we can retrieve the most relevant documents to address user queries and generate responses to regulatory inquiries.

#### **Project Goals**

- Streamline Legal Preparation
- · Reduce workload for target users by automating document retrieval

#### Improve Document Search Efficiency · Quick retrieval of relevant filings and rulings

- from large collection of stored documents
- Ensure Consistent Response Output . Generate standardized and uniform answers to regulatory inquiries across all submissions

#### Challenges

#### Document Ingestion

. Navigating the CPUC Website & processing data from thousands of proceedings

semantic and structured filtering

Search Optimization · Narrowing relevant search results combining



**System Architecture** 























## Research Updates

### **Current Outline**

Part 4, 7, and 8 done

#### 1. Introduction

- 1.1 Context and Motivation
- 1.2 Research Problem & Contributions

#### 2. Related Work

- 2.1 Expert Systems in Regulatory / Compliance Domains
- 2.2 Semantic Search & RAG with LLMs
- 2.3 Gaps GRC Addresses

#### 3. System Architecture

- 3.1 High-Level Block Diagram
- 3.2 Data Flow
- 3.3 Component Descriptions

#### 4. Implementation Details

- 4.1 Embedding pipeline & vector store
- 4.2 LLM orchestration & prompt templates
- 4.3 Back-end & front-end
- 4.4 Performance optimizations

#### 5. Experimental Setup

- 5.1 Dataset (CPUC filings used)
- 5.2 Baselines (keyword search)
- 5.3 Metrics
- 5.4 Participant details (developer lab study)

#### 6. Results

- 6.1 Quantitative results (tables/graphs)
- 6.2 Usability findings

#### 7. Discussion

- 7.1 Interpretation & practical implications
- 7.2 Limitations (lab vs. field)
- 7.3 Lessons learned

#### 8. Conclusion

Recap contributions + Next steps

- 9. Glossary
- 10. References





## **Vector Database**

### **Finalized Database**

- Stores vectors of all documents from proceedings filed prior to June 25, 2025 since
  2020
- Final set of documents is approximately 50 GB and 1,200 proceedings
  - Approximately 2.3 million vectors along with their text stored in the Qdrant database



## Web Hosting

### **SSL** Certificate

- unable to generate through letsencrypt due to firewall issues
- asking UCI for a new one

### In the meantime...

Switching to Docker + AWS



# Questions?

