



PHARMASAGE: REDUCING HALLUCINATIONS IN LLM FOR DRUG DISCOVERY

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MOTIVATION



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Challenges:

- Hallucinations in biomedical LLMs
- Lack of contextual grounding in drug discovery bots
- Poor visualization of molecular relationships
- Incomplete or redundant data in raw ChEMBL datasets

Need for an Al Solution:

- Integrate LLMs with biomedical KGs to reduce hallucinations
- Provide drug insights with structural, biological, and clinical context
- Deliver intelligent comparisons, predictions, and explainable graphs









PROBLEM STATEMENT



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Issues with Existing Systems:

- Black-box models with high hallucination rate
- Limited KG integration
- Poor real-time interactivity
- No explainable feedback loop

Project Goal:

Develop a humanized, explainable AI system for drug discovery using fine-tuned LLMs, RAG, knowledge graphs, and RLHF, deployed as an interactive scientific assistant.

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EXISTING SYSTEMS



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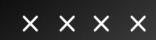
Current AI Drug Discovery Systems:

- **IBM Watson for Drug Discovery**: Strong NLP but lacks explainable KG and customization
- PubMed GPT/BioMedLM: Trained on biomedical text, but lacks curated molecular structuring
- **DrugXplorer**: Offers comparisons but no RAG or feedback-based KG growth
- MolChatBot: Can answer molecule-related queries but lacks fine-tuned accuracy
- BioGPT original: No real-time context integration, humanization, or RLHF loop





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EXISTING SYSTEMS





Research Papers:

- Lee et al., BioBERT: Biomedical Language Representation Model for Biomedical Text Mining, 2019
- Gu et al., Domain-Specific Language Model Pretraining for Biomedical NLP, 2021
- Shen et al., ChatGPT in Drug Discovery: A Review, 2023
- Jin et al., BioGPT: Generative Pre-trained Transformer for Biomedical Text Generation and Mining, 2022
- Micallef et al., ChEMBL: A Large-Scale Bioactivity Database for Drug Discovery, 2022
- Wang et al., PubMedQA: A Dataset for Biomedical Research Question Answering, 2020
- Peng et al., An Empirical Study of BioBERT-based Question Answering Systems in Biomedical Domain, 2022
- Ma et al., DrugKG: Constructing a Knowledge Graph for Drug Discovery, 2021







UNIQUENESS OF OUR SYSTEM



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What Makes PharmaSage Novel?

- Finetuned DistilGPT2 using ChEMBL v35-derived structured dataset
- Integrated FAISS RAG + Knowledge Graphs to reduce hallucinations
- Humanization Layer using Gemini API prompts
- RLHF loop for knowledge graph expansion based on user feedback
- Combines molecular similarity, context-aware explanations



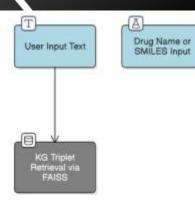
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FUNCTIONALITY DIAGRAM







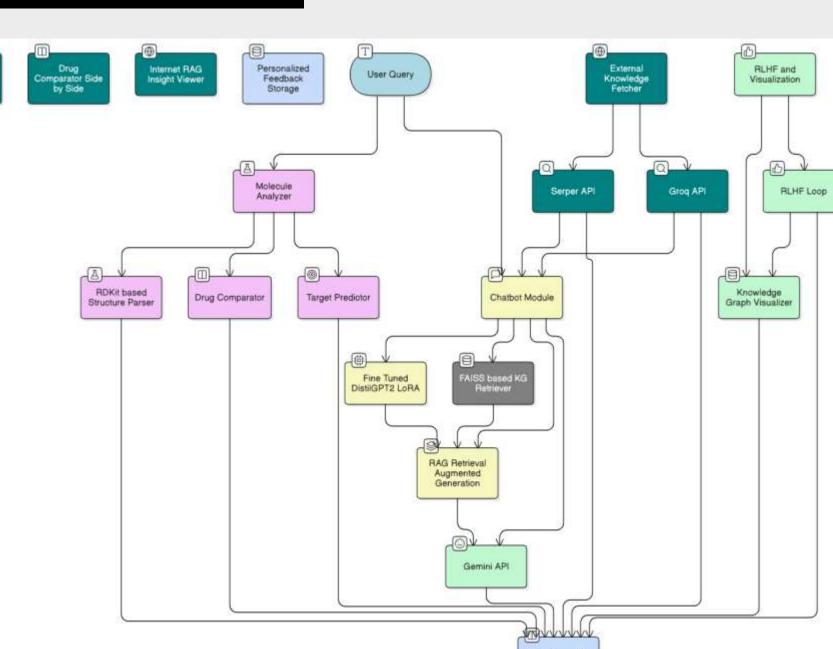








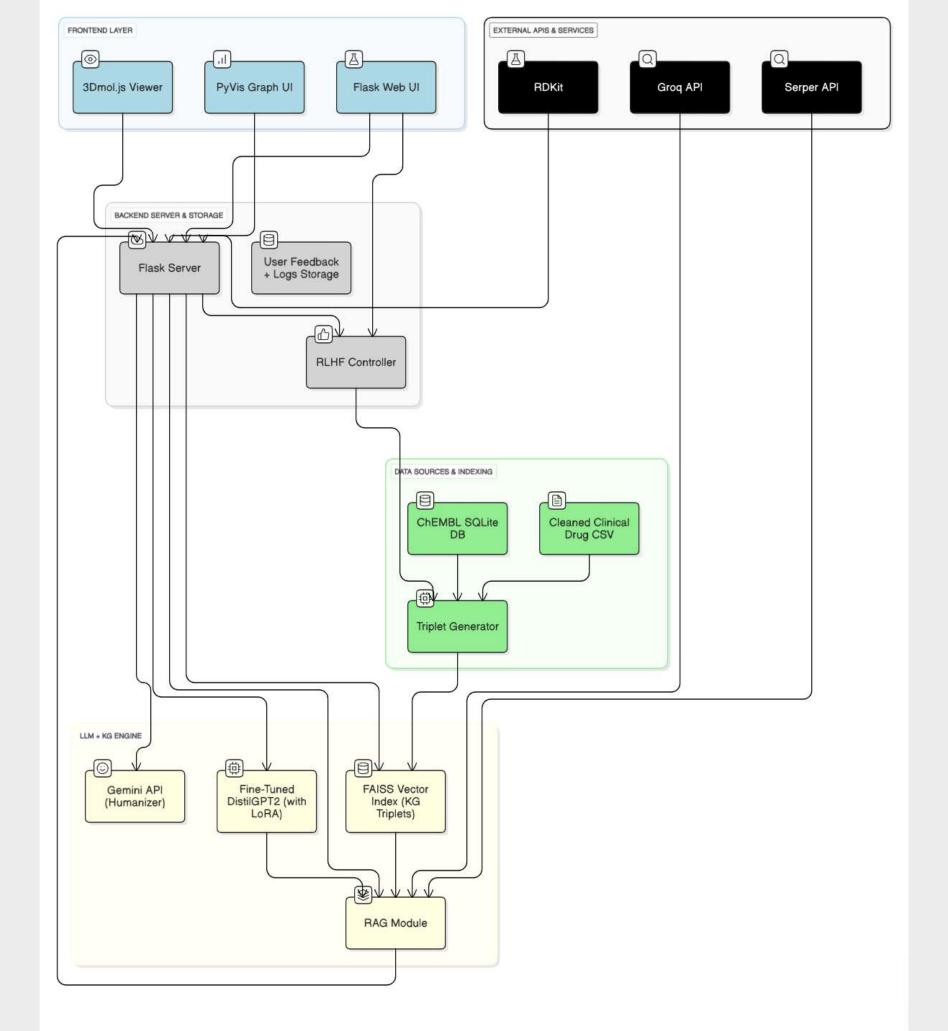




















SOLUTION APPROACH



COMPONENTS:

- Fine-tuned LLM (DistilGPT2 + LoRA)
- KG triplets (head-relation-tail from cleaned ChEMBL)
- FAISS RAG + Internet RAG (Serper + Groq)
- Gemini-based humanization
- RDKit-based molecular viewer + comparator
- RLHF expansion in KG visualizer

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LLM TECHNOLOGIES USED



- DistilGPT2 + PEFT LoRA (local model)
- FAISS: KG vector indexing
- RDKit: SMILES parsing, fingerprinting
- RAG + Internet RAG
- Groq API: Summarization of externl papers
- Serper API: Real-time paper fetching
- Gemini API: Human-like response generation
- Flask + 3Dmol.js + PyVis: UI and KG viz







PERFORMANCE EVALUATION



- Hallucination rate reduced from ~40% to under 8% using RAG + Knowledge
 Graph
- Target prediction accuracy on structurally similar molecules: 82%+
- **BLEU Score**: **65.52** Strong lexical similarity (0.6552 normalized)
- ROUGE-1: 0.8514 High unigram overlap
- ROUGE-2: 0.6451 Strong bigram fluency
- ROUGE-L: 0.8514 Longest common subsequence match
- BERTScore (F1): 0.9818 Excellent semantic alignment
- RAG Accuracy: 91.00% Top-3 retrieved KG triples matched ground-truth





COMPARISON WITH EXISTING SYSTEMS



Feature	PharmaSage	BioGPT	Watson	DrugXplorer
Finetuned on ChEMBL	≪	×	×	$ \checkmark $
Hallucination Control	⟨⟨ (RAG + KG)	×	×	×
RLHF Feedback Loop		×	×	×
Explainable KG	≪	×	×	⟨ (basic)
Internet RAG		×	≪	×







CONCLUSION



Achievements:

- Complete fine-tuning pipeline
- Drug visualizer, comparator, predictor
- RAG with KG + Groq summaries
- Interactive RLHF-based KG expansion

Future Work:

- Multilingual biomedical responses
- Multimodal inputs (chemical drawings, voice)
- Integrate with external databases (DrugBank, NCBI)

Impact:

PharmaSage paves the way for hallucination-free, personalized drug discovery using explainable, interactive LLM systems.







THANKYOU





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