

ANSWER ASSISTANT WEBSITE Mangesh Singh Siddharth Sahu **Tushar Garg**

QUESTION AND FOR PLAKSHA'S

PROBLEM STATEMENT

Navigating through Plaksha's website for specific information can be sometimes overwhelming, especially for those who are not part of the Plaksha community.

MOTIVATION

This will make fetching information from the website more convenient for users.





POTENTIAL APPLICATIONS

Question and Answer Assisstant

POTENTIAL IMPACT

Enhanced User Experience, personalised assistance, innovation and differentiation

LITERATURE SURVEY



FINETUNING LLM

Finetuning pretrained parameters: Adjusting the weights of pretrained models on a smaller, taskspecific dataset to improve performance on that task.



RETRIEVAL AUGMENTED GENERATION

Retrieval-Augmented Generation (RAG) fetches relevant data from outside the foundation model and enhances the input with this data, providing richer context to improve output.



RETRIEVAL AUGMENTED GENERATION

2. Pass the embedded query vector to our vector DB.

3. Retrieve the top-k relevant contexts measured by distance between the query embedding and all the embedded chunks in our knowledge base.



 Pass the query to the embedding model to semantically represent it as an embedded query vector.

4. Pass the query text and retrieved context text to our LLM.

5. The LLM will generate a response using the provided content.

WHY RAG **OVER OTHER RAA'S LIKE REALM**



Retrieval Augmented Architectures Retrieval-Augmented Language Model Pre-Training

> Due to its excellent open-source documentation and availability.



Less computationally expensive

due to partial end-to-end training

REFERENCES AND CITATIONS

How To Choose Perfect LLM For The Problem Statement Before Finetuning https://www.labelle rr.com/blog/howto-choose-llm-tosuit-for-use-case/

A Beginner's Guide to Large Language Models https://resources.nvidi a.com/en-us-largelanguage-modelebooks

Fine-Tuning Llama-2: A Comprehensive Case Study for Tailoring Models to **Unique Applications**

https://www.anyscale.com/ blog/fine-tuning-llama-2-acomprehensive-casestudy-for-tailoring-modelsto-unique-applications

Building RAG-based LLM Applications for Production (Part 1)

https://www.anyscale. com/blog/acomprehensiveguide-for-buildingrag-based-llmapplications-part-1

DATA COLLECTION

Considerations

- Authentic
- Accurate
- Reliable



DATA TYPE?

Text

ETHICAL ISSUES?

Not Really



DATA COLLECTION: HOW?

1.Web Scraping Tool 2.Manually

DATA COLLECTION: How much?

1. Website: 222 Data Points 2. QnA Pairs: ~ 1000



isha mission. The founders are guided by a distinguished Acade Board. What we do and how we do it is represe Pillar 02 Pillar 03 Enabling research and Addressing grand innovation ecosystem challenges Plaksha's journey started in 2015, as a conversation among a few friends about how and why technology education needs to change. This group of people, most of them entrepreneurs and business leaders, were keen to change the landscape of engineering and technology education in India and the world. Reimagin Higher Education Foundation, a Section 8 not-for-profit company, was set up in 2017 with a mission of converting the idea into a plan. In Feb 2019, the groundbreaking of Plaksha University's campus in Mohali took place with the aim of opening the campus in Aug 2021 Over 60 like-minded tech entrepreneurs, business leaders and academicians across 6 countries are a par of the Plaksha mission. The founders are guided by a distinguished Academic Advisory Board. What we do and how we do it is represented by three pillars - based on the shared inspiration of academ 3 leaders and the founding group. The founders are business leaders and technology entrepreneurs based in Hong Kong, India, London, New York, Silicon Valley, Singapore and Tokyo . They play an active role not only by investing philanthropically but also by mentoring students and building a community of change-makers and impact creators. our Founders In 2017, the founding members reached out to change-makers and radical thinkers in academia and our Academic Advisory Board was formed. Today, it has 14 eminent academicians and thought leaders from across the globe - each at the forefront of transforming education in their respective institutions. They play Our Academic Advisory Board an instrumental role guiding our vision and academic programs. Our partners share our vision - to reimagine engineering education and research. Together, we work closely We partner with top institutions, globall on curriculum design, delivery, faculty and student exchanges and collaborative research By 2035 we plan to have Students across our programs U. VERSITY

DATA PREPROCESSING

Chunking

Synonym Augmentation

Text to vector conversion

ML METHODOLOGY



Generation Models Integration Techniques

Implemented Guardrails

Domain Specific Adaptation



CHALLENGES





documentation

- Hardware: Unavailability of
- computational resources

2)Software: Lack of proper

PERFORMANCE METRICS



GOOGLE PALM

Semantic Similarity Score



Rouge-1 Score



DEPLOYABILITY CHALLENGES

Increased Latency

Resource Limitations

Model Optimization



THANK YOU FOR LISTENING!

Feedback and questions, please:)

We're grateful to Prof. Siddharth for challenging us, and pushing us to more.