

## Oops!

Welcome back! Our systems are currently free, and we're excited to continue to develop retrieval part of the RAG. Please relax and enjoy the content.



Homayoun Afshari



Arash Daneshvar



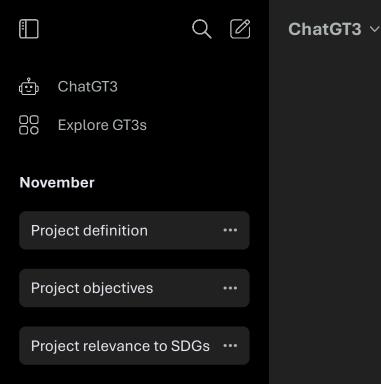
Hossein Khodadadi

### You had hit the free plan limit for GT3.

You can use GT3 or upgrade to GT3-40 to continue this chat. Your limit is lifted since it's December 4th.



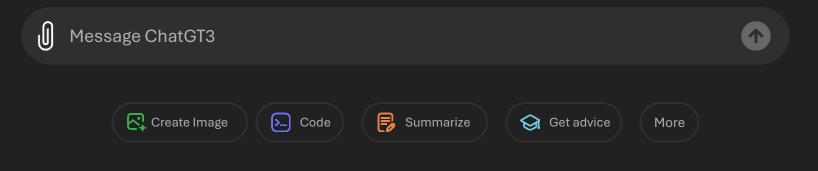
Get plus



## What can I help with?

0

?



Design: customer journey •••

Design: functional diagram •••

Manage: WBS

Manage: Gantt

Conclusion

00 Explore GT3s

### November

Project definition

Project objectives

Project relevance to SDGs ...

Design: user personas

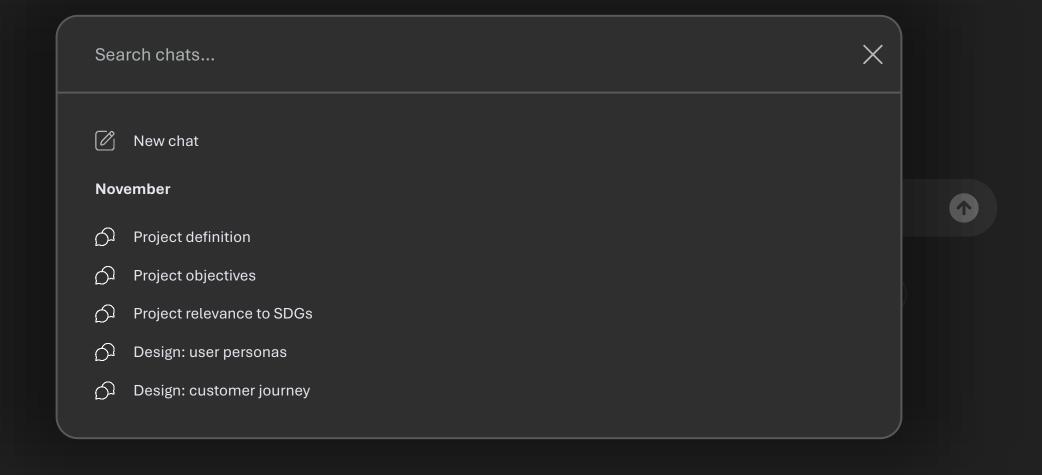
Design: customer journey •••

Design: functional diagram •••

Manage: WBS

Manage: Gantt

Conclusion



ChatGT3

000 Explore GT3s

### November

Project definition

Project objectives

Project relevance to SDGs ...

Design: user personas

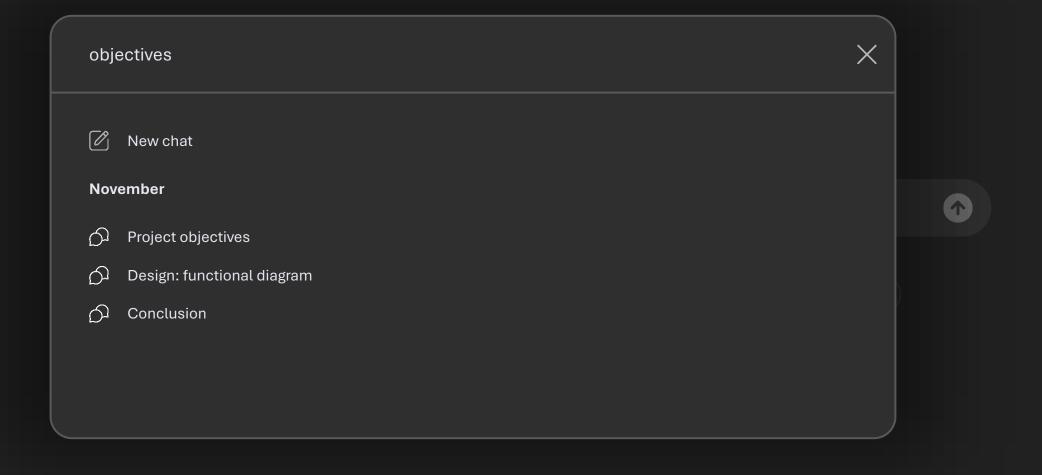
Design: customer journey •••

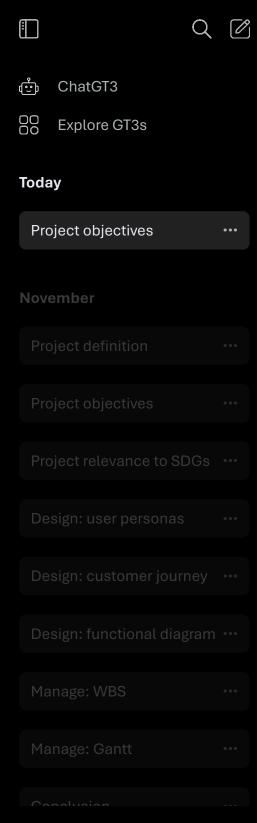
Design: functional diagram •••

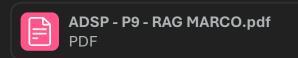
Manage: WBS

Manage: Gantt

Conclusion







Please describe the objectives of the project explained in this file.

ChatGT3 ~

Let's dive into optimizing information quality in a RAG system, with a spotlight on retrieval and response generation.

When it comes down to it, the quality of information is everything.

Based on insights from the PDF, here's a breakdown of the main objectives:

- **Objective 1**: Enhance the Retrieval Process Using LLMs
- Objective 2: Elevate the Quality of Context for the Response Generator

And here's a potential **research question** to dig deeper:

"How optimizing the retrieval stage in a RAG system ensures high-quality and contextually accurate responses?"

Also, for this project, the **value proposition** could be:

"Boosting Accuracy through Advanced Retrieval"

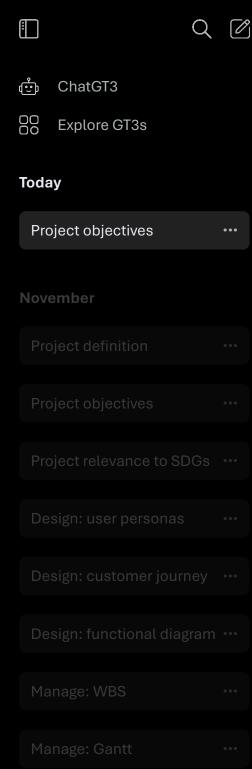
This approach keeps the focus sharp - enhancing accurate retrieval and delivering relevant.

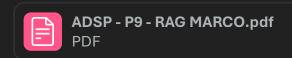


Message ChatGT3



Upgrade plan





Please describe the objectives of the project explained in this file.

ChatGT3 ~

Let's dive into optimizing information quality in a RAG system, with a spotlight on retrieval and response generation.

When it comes down to it, the quality of information is everything.

Based on insights from the PDF, here's a breakdown of the main objectives:

- **Objective 1**: Enhance the Retrieval Process Using LLMs
- Objective 2: Elevate the Quality of Context for the Response Generator

And here's a potential **research question** to dig deeper:

"How optimizing the retrieval stage in a RAG system ensures high-quality and contextually accurate responses?"

Also, for this project, the value proposition could be:

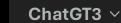
"Boosting Accuracy through Advanced Retrieval"

This approach keeps the focus sharp - enhancing accurate retrieval and delivering relevant.



Great! Could you provide us with the updated objective for the project? |











Explore GT3s

### **Today**

**Project objectives** 

ADSP - P9 - RAG MARCO.pdf

Great! Could you provide us with the updated objective for the project?

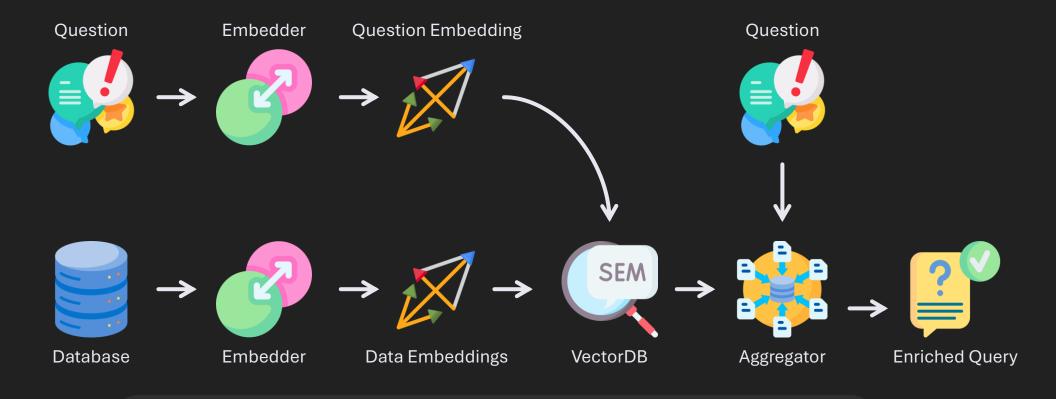


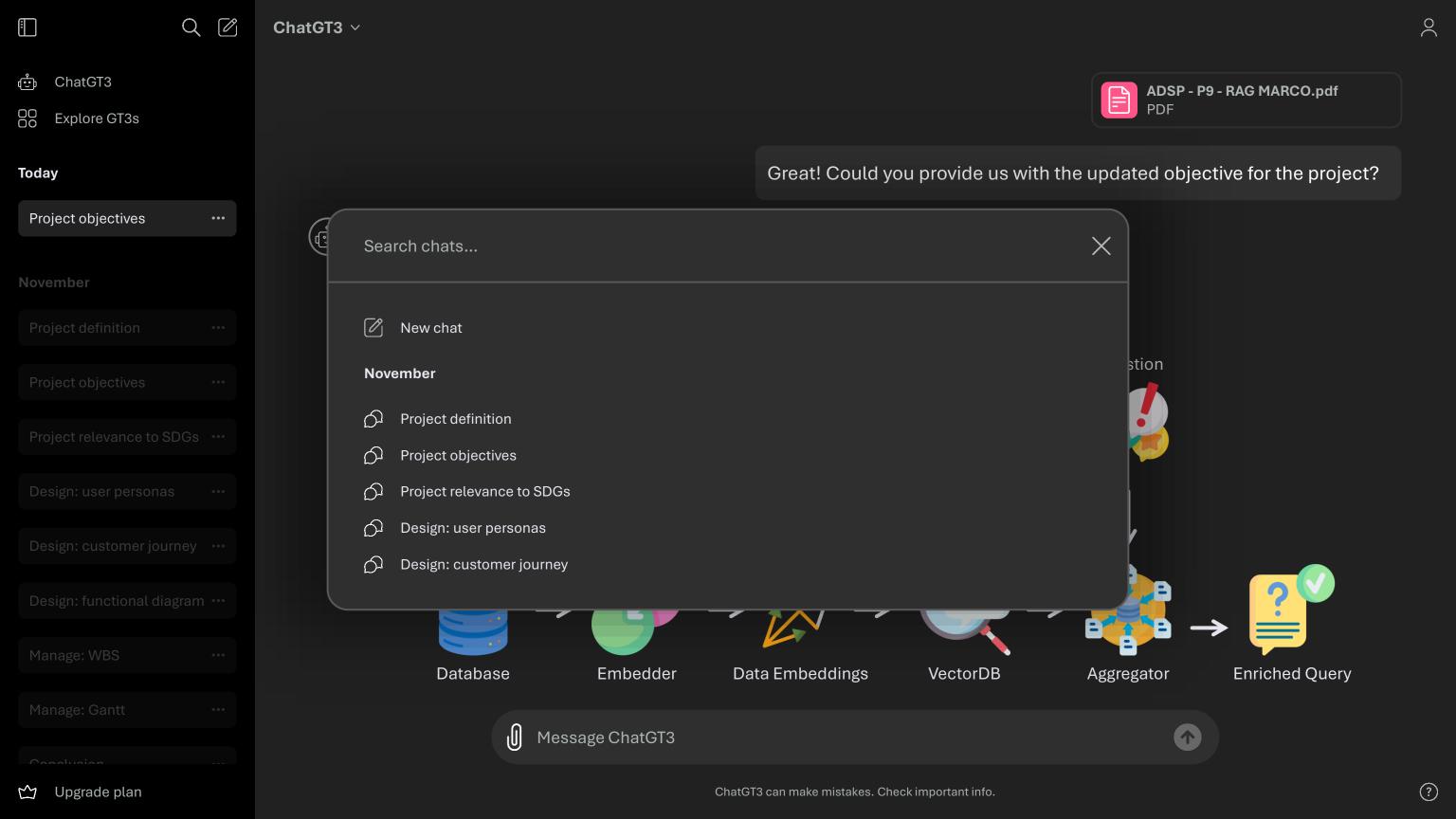
Certainly! Building on the previous objectives, here is a refined breakdown of the updated objectives of the project:

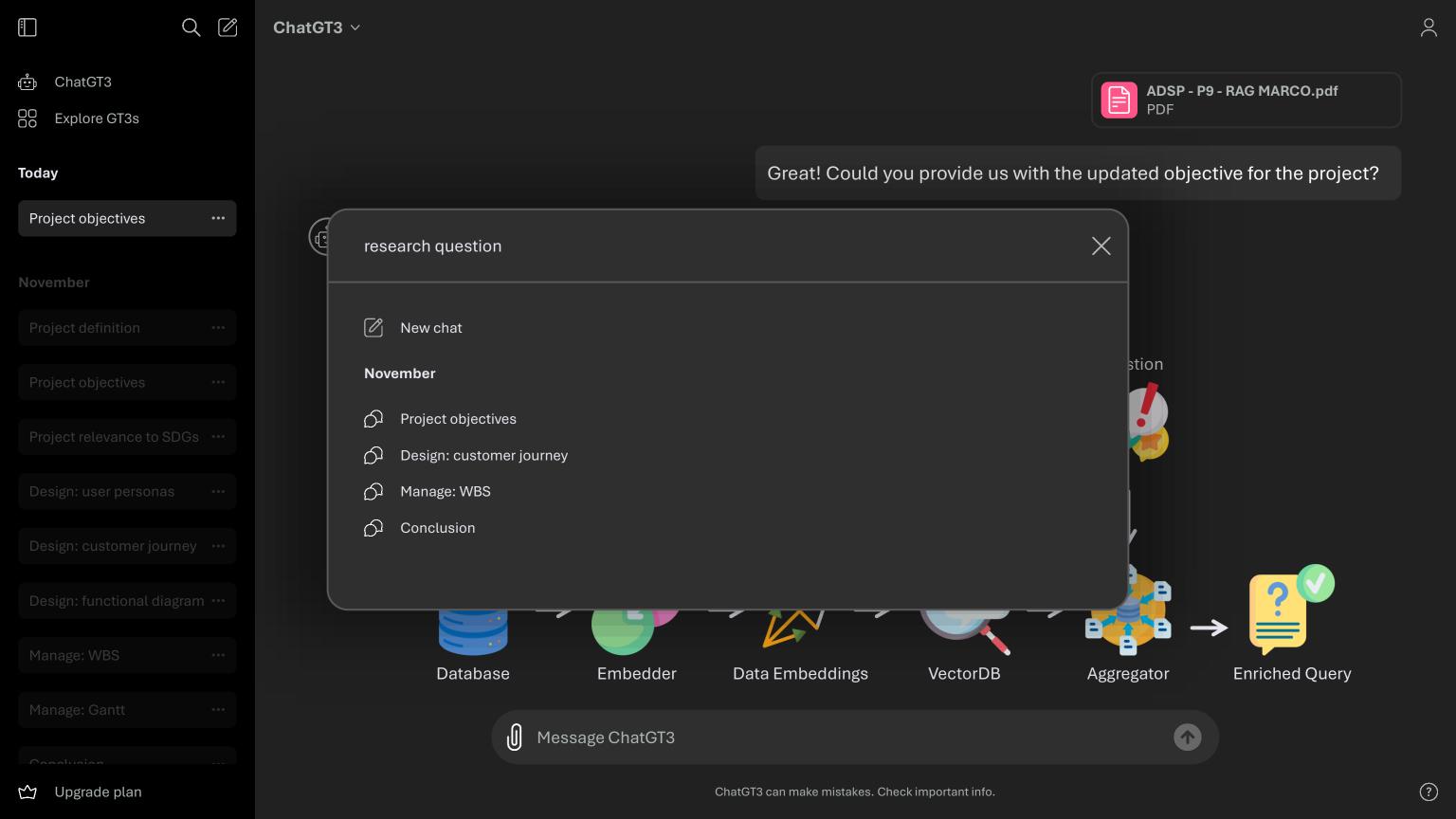
**Objective 1**: Refine the retrieval system.

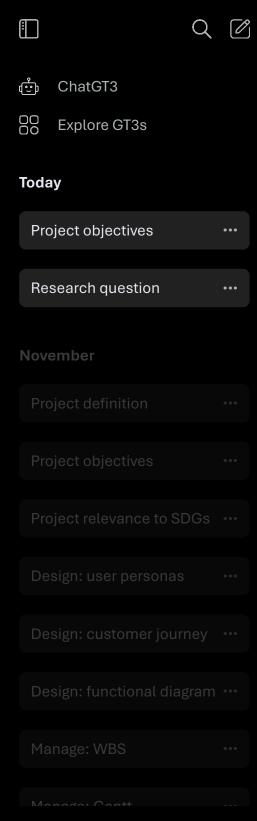
Message ChatGT3

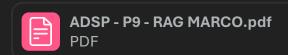
Objective 2: Improve the contextual quality of information fed into the response generation stage.











Please describe the objectives of the project explained in this file.

ChatGT3 ~

Let's dive into optimizing information quality in a RAG system, with a spotlight on retrieval and response generation.

When it comes down to it, the quality of information is everything.

Based on insights from the PDF, here's a breakdown of the main objectives:

- Objective 1: Enhance the Retrieval Process Using LLMs
- Objective 2: Elevate the Quality of Context for the Response Generator

And here's a potential **research question** to dig deeper:

"How optimizing the retrieval stage in a RAG system ensures high-quality and contextually accurate responses?"

Also, for this project, the **value proposition** could be:

"Boosting Accuracy through Advanced Retrieval"

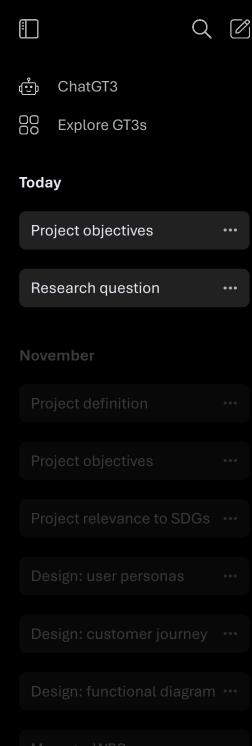
This approach keeps the focus sharp - enhancing accurate retrieval and delivering relevant.

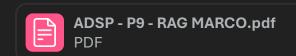


Message ChatGT3



Upgrade plan





Please describe the objectives of the project explained in this file.

L C

ChatGT3 ~

Let's dive into optimizing information quality in a RAG system, with a spotlight on retrieval and response generation.

When it comes down to it, the quality of information is everything.

Based on insights from the PDF, here's a breakdown of the main objectives:

- Objective 1: Enhance the Retrieval Process Using LLMs
- Objective 2: Elevate the Quality of Context for the Response Generator

And here's a potential **research question** to dig deeper:

"How optimizing the retrieval stage in a RAG system ensures high-quality and contextually accurate responses?"

Also, for this project, the value proposition could be:

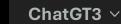
"Boosting Accuracy through Advanced Retrieval"

This approach keeps the focus sharp - enhancing accurate retrieval and delivering relevant.

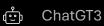


Great! please provide an updated version with a focus on the new direction.





ADSP - P9 - RAG MARCO.pdf



Explore GT3s

### **Today**

**Project objectives** 

Research question

Upgrade plan

Great! please provide an updated version with a focus on the new direction.

Now you can view two options: text-based and visualized responses.

Response 1

Here's a potential **research question** to dig deeper:

"How optimizing the retrieval stage in a RAG system ensures high-quality and contextually accurate responses?"

And the value proposition:

"Enhanced Retrieval, Delivering Smarter Responses"

Aligned with the following user persona:

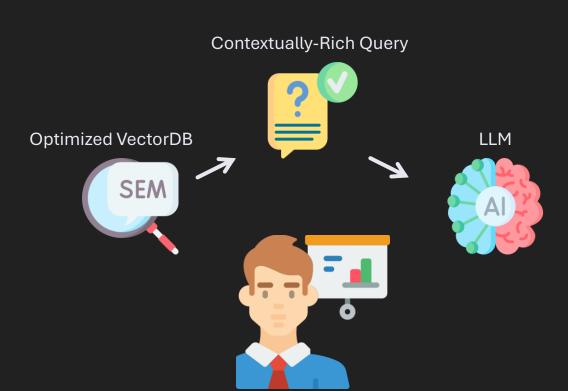
"Michael Foster, the Al software manager"

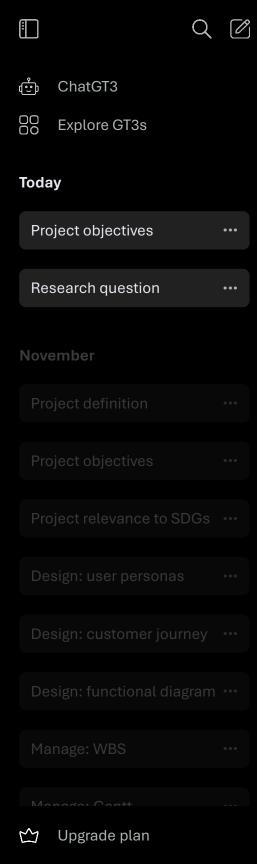


Message ChatGT3

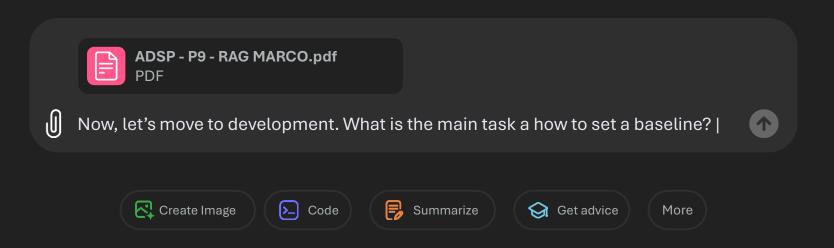


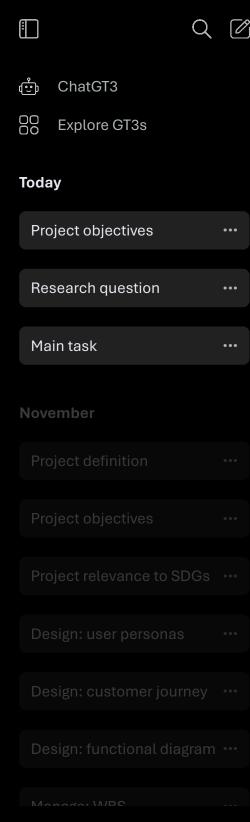
The visualization is as follows:

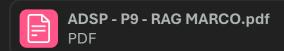




# What can I help with?







Now, let's move to development. What is the main task a how to set a baseline?



ChatGT3 ~

The main task of this phase involves several key decisions as follows:

- Choosing Datasets: Select appropriate datasets for training and evaluating the retrieval system.
- Analyzing Useful Sentence Embedders: Identify models that effectively convert text into vector representations.
- Analyzing Useful VectorDBs: Explore databases optimized for storing and querying vector embeddings.
- Selecting Similarity Measures: Choose methods to compare embeddings for accurate retrieval.
- Selecting Performance Metrics: Determine metrics to evaluate and benchmark system performance.
- Forming A Baseline: Develop an initial system configuration to measure performance and guide improvements.
- Making Improvements on Top of Baseline: Iteratively refine the system by testing enhancements to improve the
  retrieval quality and achieve the defined objectives.

A well-defined baseline is crucial as it provides a reference point to measure progress and evaluate improvements. It ensures that enhancements are meaningful by establishing a solid foundation for comparison and identifying areas of strength or weakness.

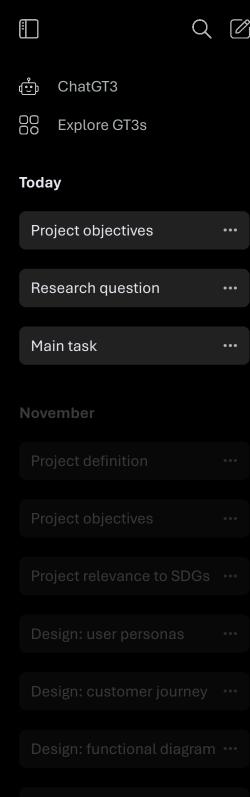
Would you like me to also explain the baseline in the same manner as that of the main task?

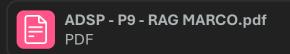


Message ChatGT3



Upgrade plan





Now, let's move to development. What is the main task a how to set a baseline?



ChatGT3 ~

The main task of this phase involves several key decisions as follows:

- Choosing Datasets: Select appropriate datasets for training and evaluating the retrieval system.
- Analyzing Useful Sentence Embedders: Identify models that effectively convert text into vector representations.
- Analyzing Useful VectorDBs: Explore databases optimized for storing and querying vector embeddings.
- Selecting Similarity Measures: Choose methods to compare embeddings for accurate retrieval.
- Selecting Performance Metrics: Determine metrics to evaluate and benchmark system performance.
- Forming A Baseline: Develop an initial system configuration to measure performance and guide improvements.
- Making Improvements on Top of Baseline: Iteratively refine the system by testing enhancements to improve the
  retrieval quality and achieve the defined objectives.

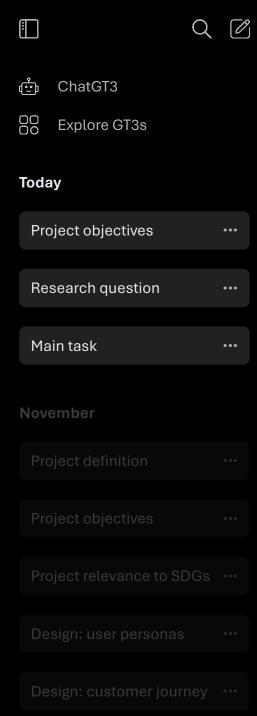
A well-defined baseline is crucial as it provides a reference point to measure progress and evaluate improvements. It ensures that enhancements are meaningful by establishing a solid foundation for comparison and identifying areas of strength or weakness.

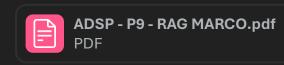
Would you like me to also explain the baseline in the same manner as that of the main task?



Yes, please also explain the baseline.

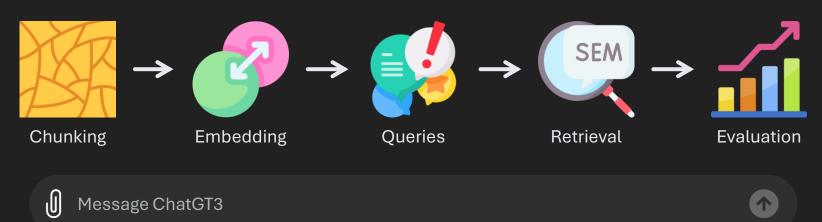


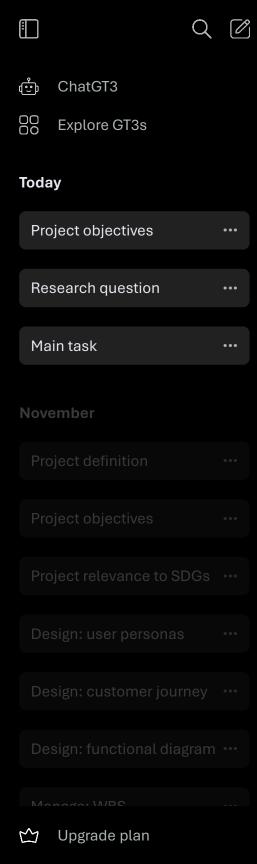




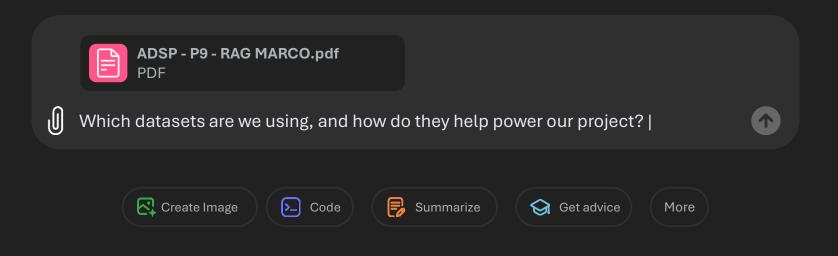
Yes, please also explain the baseline.

- **The baseline** for the system involves the following steps:
  - **Document Chunking:** Break documents into smaller chunks to improve retrieval granularity.
  - Embedding Chunks and Storing in VectorDB: Convert chunks into vector representations and store them in a
    - Taking A Random Sample from Queries and Obtaining Their Embeddings: Start with a manageable set of
  - Top-K Similar Passage Retrieval for Each Query: Dynamically retrieve the top-k most relevant passages for each
  - **Performance Evaluation**: Assess the retrieval system's effectiveness using predefined performance metrics.





# What can I help with?





ChatGT3

Explore GT3s

### **Today**

**Project objectives** 

Research question

Main task

Datasets

Upgrade plan



Which datasets are we using, and how do they help power our project?

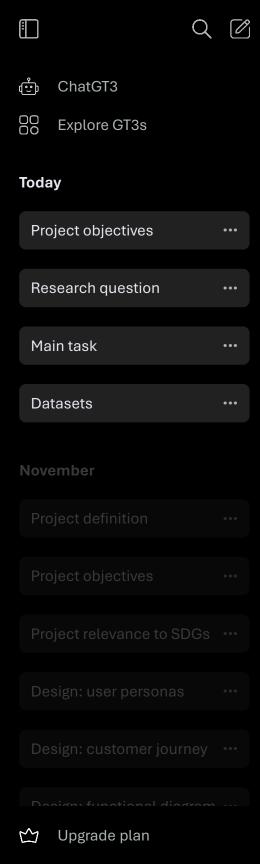


### MS-MARCO and TriviaQA.

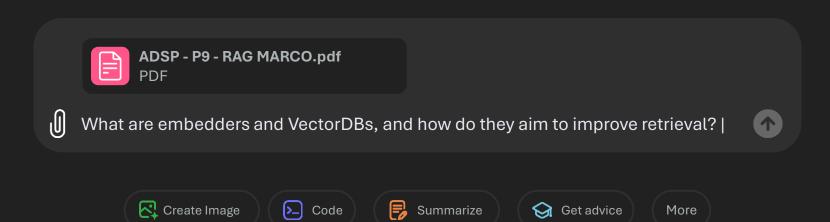
Datasets	Question Format	Relevant Passages	Selected Passages	Passage Ranking	Generated Answer
MS-MARCO	Query strings asked by users on Bing.	Approximately six passages related to the query.	Approximately two passages selected after retrieval,	No explicit ranking.	A well-formed human-generated answer.
TriviaQA	Trivia questions collected from quiz-league websites.	Approximately six passages containing the answer.	Same as the retrieved passages.	Ranked based on similarity to the question.	Human-verified and machine- generated subsets

Message ChatGT3





# What can I help with?



 $Q \mathcal{D}$ 

ChatGT3

Explore GT3s

### **Today**

**Project objectives** 

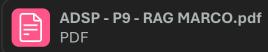
Research question

Main task

Datasets

Embedders and VectorDBs •••

ChatGT3 ~



What are embedders and VectorDBs, and how do they aim to improve retrieval?



all-mpnet-base-v2, multi-qa-mpnet-base-dot-v1, and all-distilroberta-v1.

Embedders	Architecture	Use Case	Performance
all-mpnet-base- v2	MPNet	General-purpose.	Slow with excellent embedding power
multi-qa-mpnet- base-dot-v1	MPNet	Optimized for semantic search.	Moderate with very good embedding power
all-distilroberta- v1	DistilRoBERTa	General-purpose.	Fast with good embedding power



Message ChatGT3





ChatGT3

Explore GT3s

### **Today**

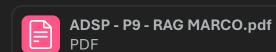
Project objectives

Research question

Main task

Datasets

Embedders and VectorDBs ···



What are embedders and VectorDBs, and how do they aim to improve retrieval?



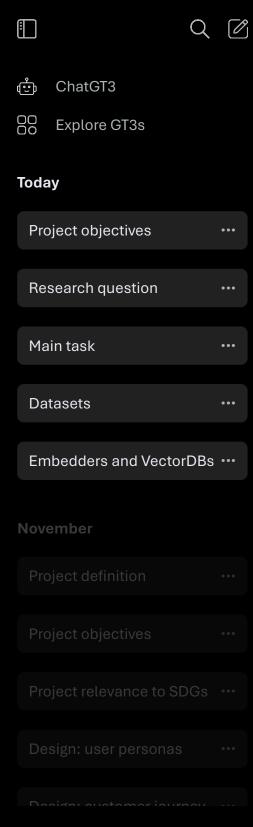
### FAISS and ScaNN.

VectorDB	Algorithm	Customizability	Optimization
FAISS (Facebook AI Similarity Search)	IVF, PQ, HNSW, and Flat indexing.	Tunable balance between accuracy and speed.	GPU acceleration and memory efficiency
ScaNN (Scalable Nearest Neighbors)	Tree-based, ADQ, and Re-ranking.	Focuses on maintaining high accuracy with speed.	CPU-focused optimizations with scalability

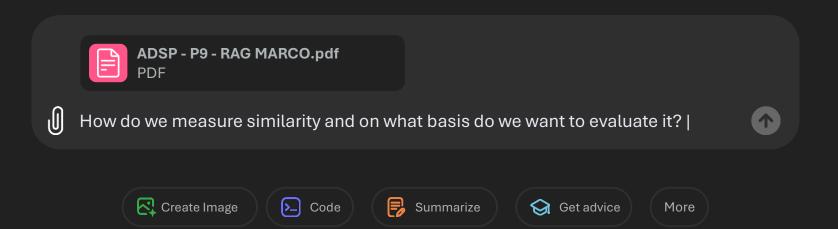


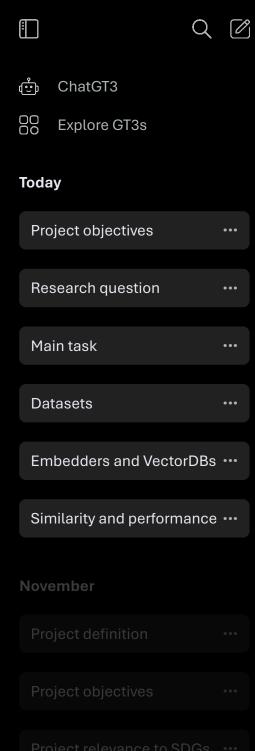
Message ChatGT3

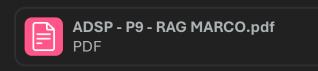




# What can I help with?

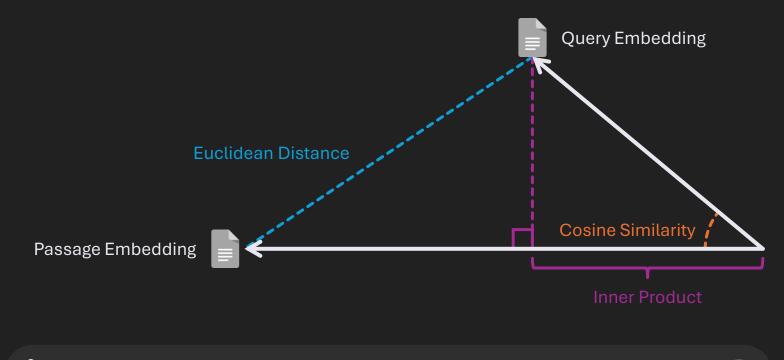




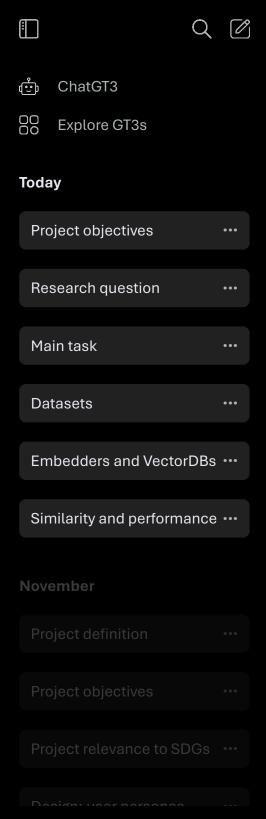


How do we measure similarity, and why is performance so crucial for this system?

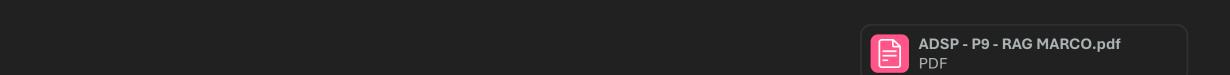
### L2 distance, Inner product, and Cosine Similarity,





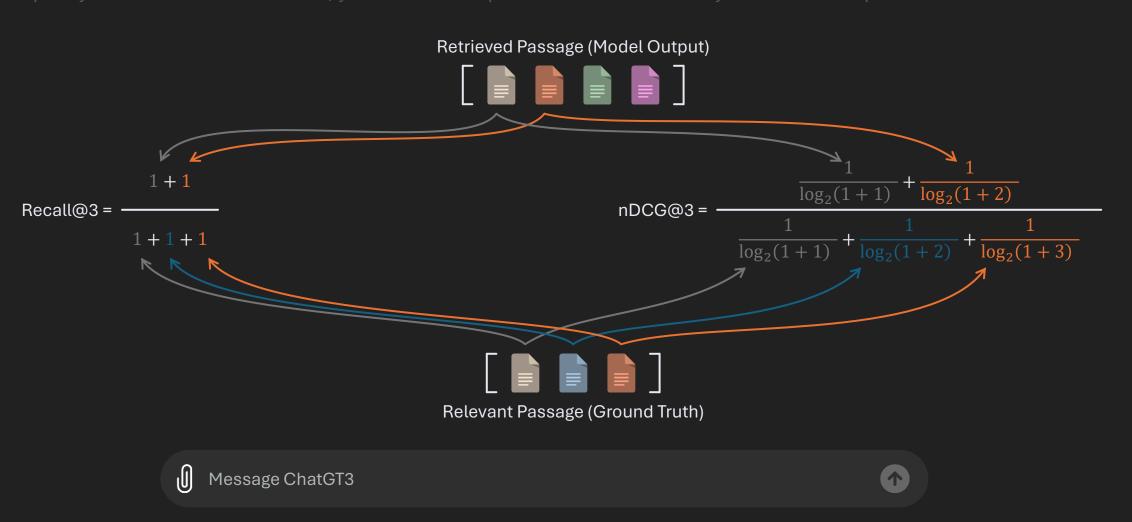


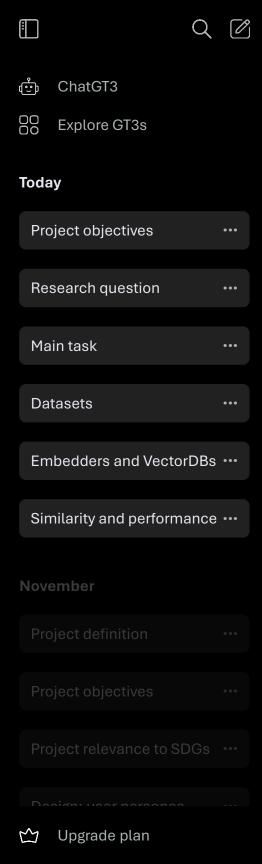
٥



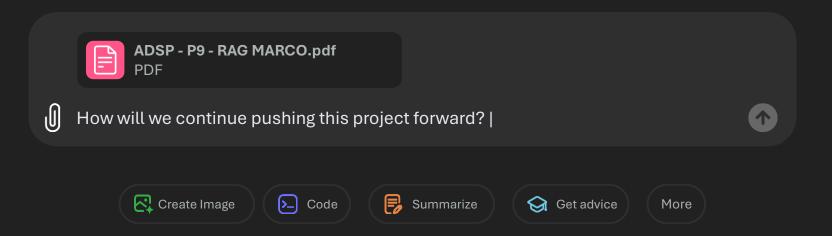
How do we measure similarity, and why is performance so crucial for this system?

For evaluating system performance, we use metrics such as **Recall@k** and **nDCG@k**, ensuring relevance and ranking quality in retrieved results. Below, you'll find a comparison of these similarity measures and performance metrics.

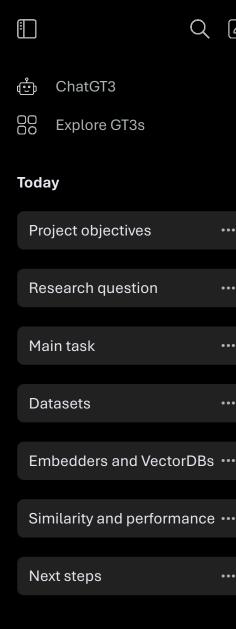


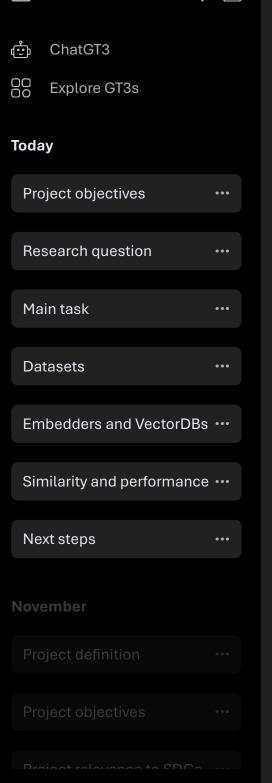


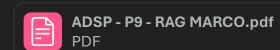
## What can I help with?



ChatGT3 ~





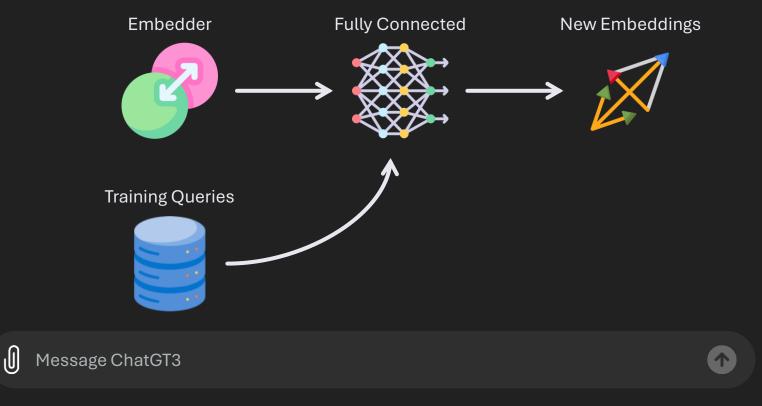


How will we continue pushing this project forward?

ChatGT3 ~

Applying a fully connected layer to the current embedder models and training it using the available datasets.

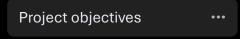
- Enriching the embeddings to bring query and document embeddings closer together in the embedding space,
- Reducing the dimensionality of the representations below the original value to provide compression capability,



Explore GT3s

How will we continue pushing this project forward?

### **Today**



- Research question
- Main task
- **Datasets**
- Embedders and VectorDBs ···
- Similarity and performance •••
- Next steps

### November



On top of the semantic similarity scores measured by the fully connected architecture, we can also employ NER,

Keyword, and Topic extraction to obtain **syntactic similarity scores** through applying the following measures:

- **NER Extraction**: Identify the named entities such as persons, organizations, locations, or dates, which can be
- **Keyword Extraction**: Extract specific keywords using YAKE library.
- **Topic Extraction**: Determine the main theme of passages and queries.

$$S = w_1 S_{FC} + w_2 S_{NER} + w_2 S_{KEYWORD} + w_4 S_{TOPIC}$$

Where, for the weights, we have:

$$w_1 + w_2 + w_3 + w_4 = 1$$

Also, for the components scores, we have:

$$S_{NER} = rac{\left|\mathbb{E}_Q \cap \mathbb{E}_P
ight|}{\left|\mathbb{E}_Q
ight|}$$
 ,  $S_{KRYWORD} = rac{\left|\mathbb{K}_Q \cap \mathbb{K}_P
ight|}{\left|\mathbb{K}_Q
ight|}$  ,  $S_{TOPIC} = egin{cases} 0 & t_Q 
eq t_P \\ 1 & t_Q = t_P \end{cases}$ 



Message ChatGT3





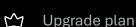
ChatGT3

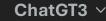
Explore GT3s

### **Today**

- **Project objectives**
- Research question
- Main task
- **Datasets**
- Embedders and VectorDBs ···
- Similarity and performance •••
- Next steps

November







How will we continue pushing this project forward?



Moreover, instead of manually tuning the weights,  $w_1$  to  $w_4$ , we can utilize a transformer architecture to learn them dynamically as attention scores. The process of this attention block is as follows:

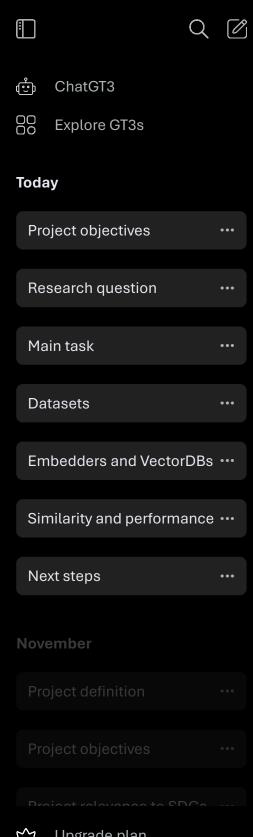
- **Keys** (*K*): The retrieved passages.
- **Queries** (*Q*): The input queries.
- **Values (V)**: The similarity scores computed for each passage  $(S_{FC}, S_{NER}, S_{KEYWORD}, \text{ and } S_{TOPIC})$ .

Note that, in the attention mechanism, the transformer computes attention scores that represent the relative importance of the similarity components, which correspond to  $w_1$  to  $w_4$  according the following formula:

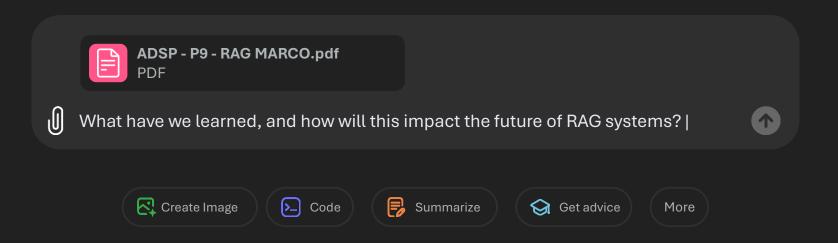
$$\begin{bmatrix} w_1 \\ w_2 \\ w_3 \\ w_4 \end{bmatrix} = \operatorname{softmax} \left( \frac{Q \times K^T}{\sqrt{d_k}} \right) V$$

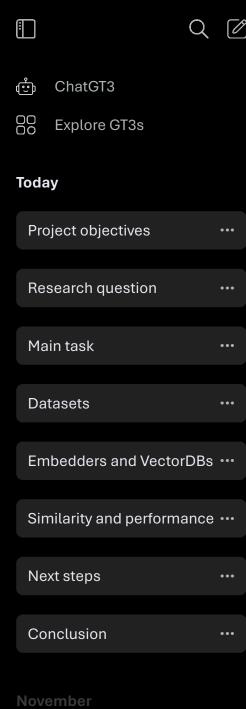


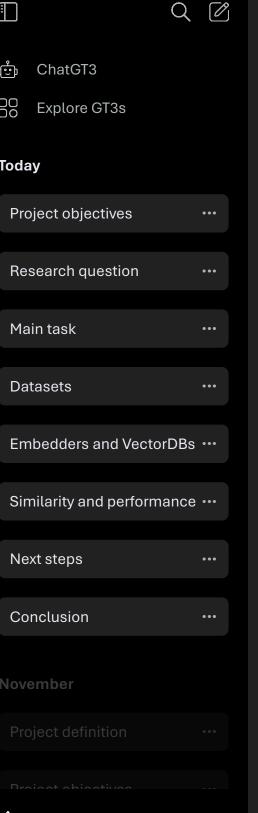


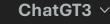


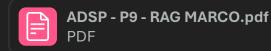
# What can I help with?











What have we learned, and how will this impact the future of RAG systems?



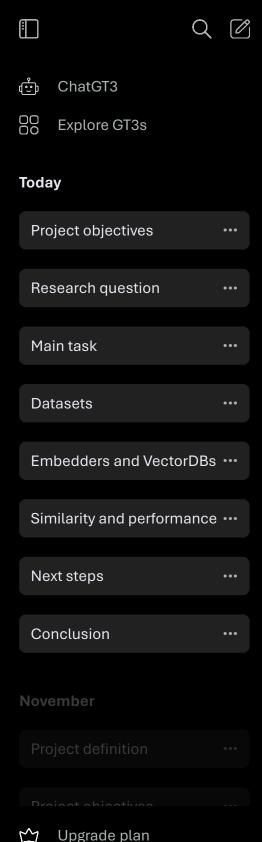
- Our work is focused on **enhancing the quality of retrieval in RAG systems** by leveraging a combination of techniques to better align query embeddings with relevant passages.
- We began by introducing the main task and established a solid baseline, utilizing vector databases, similarity measures, and key performance metrics to evaluate retrieval quality.
- We have proposed a number of strategies to take advantage of both syntactic and semantic similarities while also utilizing attention mechanism.
- Through the integration of deep learning, natural language processing, and traditional information retrieval techniques, we are hopeful that they will lead to improvements in the performance metrics utilized in this project.

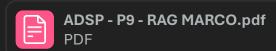


Message ChatGT3

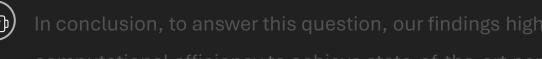


(?)





What have we learned, and how will this impact the future of RAG systems?



- Our work is focused on enhancing the quality of retrieval in RAG systems by leveraging a combination of techniques to better align query embeddings with relevant passages.
- We began by introducing the main task and established a solid baseline, utilizing vector databases, similarity measures, and key performance metrics to evaluate retrieval quality.
- We have proposed a number of strategies to take advantage of both syntactic and semantic similarities while also utilizing attention mechanism.
- follow looming the time loom retrieval Through the into You've hit the Free plan limit for GT3. techniques zed in this project. Get Plus You need GT3-40 to continue this chat because there's an attachment. Your limit resets after January 9th.



Message ChatGT3

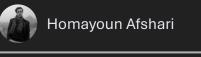


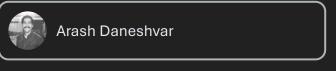


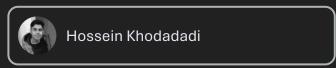
## Oops!

Thank you for your attention, but our systems are busy at the moment.

Please take a break and ask your questions.







### You've hit the Free plan limit for GT3.

You need GT3-40 to continue this chat because there's an attachment. Your limit resets after January 9th.

Get Plus