

KESHAV MEMORIAL INSTITUTE OF TECHNOLOGY

AN AUTONOMOUS INSTITUTION- ACCREDITED BY NAAC WITH 'A' GRADE Narayanaguda, Hyderabad.

Project Team No.: 33| 4001 |Research Paper Summarization

Team Members

1. Akshaya Batharaju : 23BD1A6646

2. Garige Yashasree : 23BD1A664P

3. Varshini Gopaldas :23BD1A664Q

4. Sampangi Vaishnavi :24BD5A6612

5. Lalitha Vaishnavi :23BD1A058K

Mentors

- 1. Ms.Kamal Vijetha
- 2. Mr. Shankar

CONTENTS

- 1. Title Slide
- 2. Content
- 3. Introduction
- 4. Requirement
- 5. Design
- 6. Development
- 7. Project Outcome
- 8. Conclusion
- 9. Thank you!



Research Paper Summarization

Using RAG MODEL



Problem Statement:

• Reading and understanding lengthy research papers is time-consuming. Students and researchers often need quick summaries and answers to specific questions.

End User Requirement:

- •Quick, accurate, and easy-to-read summaries of lengthy and technical research papers.
- •User must be able to choose whether he/she needs short, medium or lengthy summaries based on their time span.
- •User must be able to search for the meaning of the word that he finds difficult.
- •He must get accurate answers for the questions he poses.

Abstract:

This project focuses on developing an AI-based web application that generates a detailed summary of PDF documents using Large Language Model(LLM) and RAG Model. The main objective is to help users quickly understand lengthy documents without reading them completely. The project uses Python (FastAPI) for backend, Machine Learning models like google gemini pro for text summarization, and ReactJS for frontend. The solution enhances user productivity, especially for students and researchers handling large amounts of content.

Objectives:

- •Reduce time spent on initial research review.
- •Enhance understanding of complex topics through concise summaries.
- •Offer tailored summarization options (short, medium, long).
- •Simplify exploration of unknown terms via integrated tools.

Technology Stack:

•Artificial Intelligence & Machine Learning
FIASS as Vector Database
Retrieval-Augmented Generation (RAG)
Large Language Models (LLMs) – Google Gemini Pro



Outcome:

A web-based, intelligent summarization tool for academic content that simplifies and accelerates the research process.

Key Features:

•Automated summarization of uploaded research papers (PDF, DOC):

Enables users to upload various document formats and receive instant, Al-generated summaries.

•Adjustable summary length (short / medium / detailed):

Allows users to choose the depth of the summary based on their needs—quick scan or deep understanding.

•Dark and light mode themes:

Enhances user comfort and accessibility through customizable display preferences.

•Built-in academic dictionary:

Offers clear definitions of technical terms to aid comprehension, especially for interdisciplinary readers.

User feedback submission for improvement:

Collects input from users to refine summaries and enhance future platform updates.

•Summarization history :

History of the summaries is also stored..

•Al-powered chatbot for Q&A based on uploaded content:

Provides interactive assistance and answers user queries derived directly from the research paper content.

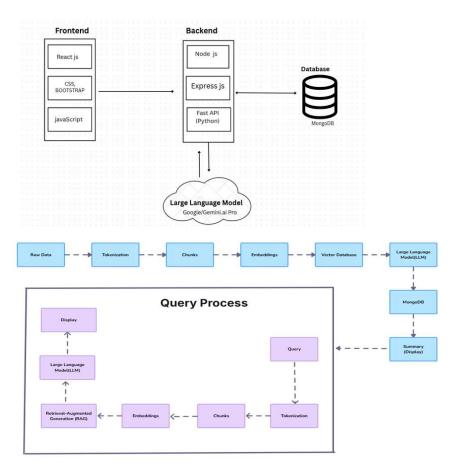
•Text to Speech :

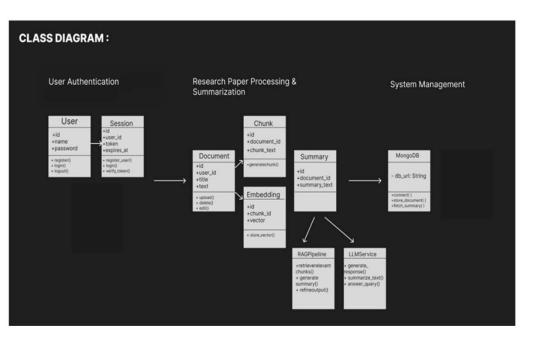
The generated summary is converted to speech.



Use Case: Large Language Model (LLM)

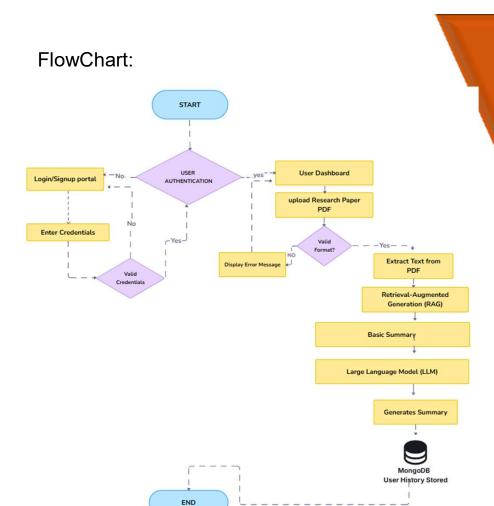
Architecture:





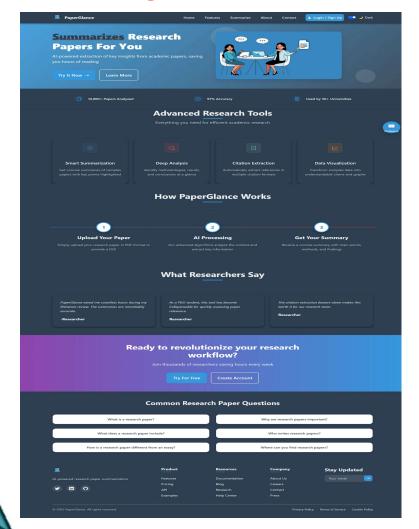
DataBase Design:

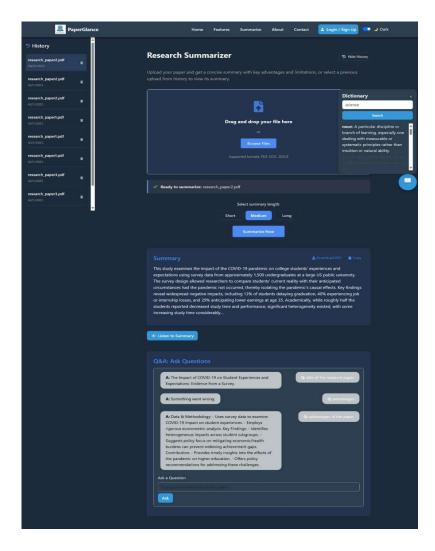




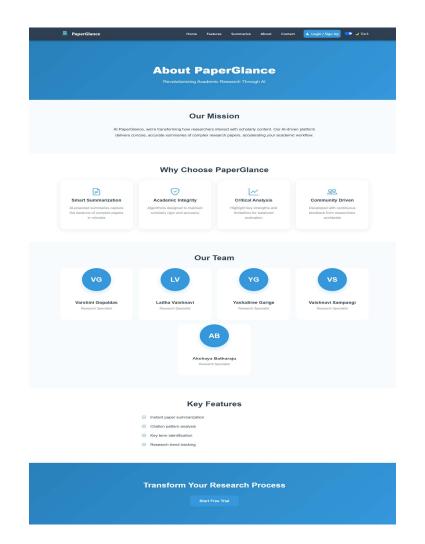


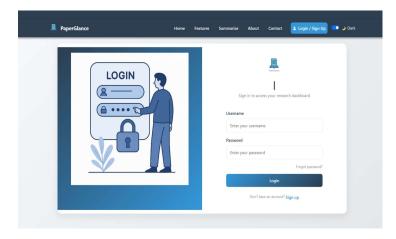
DEVELOPMENT

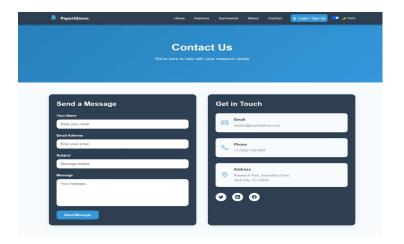














Tools and Frameworks Used:

•NLP & ML: Google/Gemini pro 1.5

•Frontend: React, Bootstrap, CSS Modules

Backend: FastAPIDatabase: MongoDB

•Others: Git, Postman, Google Colab, Figma (for UI

prototypes)

Workflow:

```
Start
IF Login/Signup → Enter credentials
 Are credentials valid?
 \rightarrow No \rightarrow Show error \rightarrow Back to login
 \rightarrow Yes
User Dashboard is displayed
User uploads a research paper (PDF / DOC)
Is the file format valid?
 \rightarrow No \rightarrow Show error \rightarrow Back to dashboard
 \rightarrow Yes
Extract text from uploaded paper
Pass extracted text to RAG (Retrieval-Augmented Generation)
Generate basic summary
Refine using Large Language Model (LLM)
Display final summary to user
Store summary and user interaction in MongoDB (user history)
User can log out / continue with another upload
  End
```



api.js

LOG error

Frontend Pseudo code

SummarizePage.js

SET API_URL = "http://localhost:8000" FUNCTION uploadFile(file): CREATE formData with file TRY: POST to API_URL + "/upload" with formData RETURN response.data CATCH: LOG error THROW "File upload failed" FUNCTION getSummary(query): TRY: POST to API_URL + "/summarize" with query RETURN response.data CATCH:

THROW "Summary generation failed"

```
// Handle upload
FUNCTION SummarizePage():
                                                                FUNCTION handleUpload(file):
                                                                  CALL API("POST", "http://localhost:8000/upload/", NEW
  SET darkMode = GET_DARK_MODE_STATUS()
                                                              FormData().append("file", file))
  SET file = NULL
                                                                  ADD response.data TO historyItems
  SET summaryResult = NULL
  SFT error = ""
                                                                // Handle summarize
  SET historyltems = EMPTY LIST
                                                                FUNCTION handleSummarize():
  SET selectedHistoryItem = NULL
                                                                  CALL API("POST", "http://localhost:8000/summarize", NEW
                                                              FormData().append("file", file))
  // Fetch history on mount
                                                                  SET summaryResult = {summary: response.data.summary}
  WHEN COMPONENT MOUNTS:
    SET historyItems = CALL API("GET",
                                                                // Render UI
"http://localhost:8000/summaries/")
                                                                DISPLAY layout WITH:
                                                                  IF historyltems:
  // Handle history item click
                                                                    FOR EACH item IN historyItems:
  FUNCTION handleHistoryItemClick(item):
                                                                       DISPLAY item.name
    SET selectedHistoryItem = item
                                                                       ON CLICK: CALL handleHistoryItemClick(item)
    SET summaryResult = {summary: item.summary}
                                                                   IF NOT selectedHistoryItem:
    CLEAR file
                                                                     DISPLAY drag-and-drop area
                                                                     ON DROP: CALL handleDrop
  // Handle download summary
                                                                     DISPLAY file input
  FUNCTION handleDownloadSummary():
                                                                     ON CHANGE: CALL handleFileChange
    CALL API("POST", "http://localhost:8000/download-
                                                                     DISPLAY summarize button
summary/", {summary text: summaryResult.summary},
                                                                     ON CLICK: CALL handleSummarize
{responseType: 'blob'})
                                                                   IF summaryResult:
    TRIGGER download OF "research summary.pdf"
                                                                     DISPLAY summary
                                                                     DISPLAY download button
  // Handle file drop/upload
                                                                     ON CLICK: CALL handleDownloadSummary
  FUNCTION handleDrop(event):
                                                                    IF selectedHistoryItem:
    SET file = FIRST event.dataTransfer.files
                                                                       DISPLAY chat history
    CALL handleUpload(file)
                                                                     ELSE:
                                                                       DISPLAY chat box
  FUNCTION handleFileChange(event):
                                                                  IF error: DISPLAY error
    SET file = FIRST event.target.files
    CALL handleUpload(file)
                                                                RETURN rendered component
```



SET logging TO DEBUG

Backend Pseudo code

backeria i seado coa

Ilm_utils.py

```
FOR EACH api key IN USED KEYS:
SET MODEL = "gemini-1.5-pro"
SET USED KEYS = CIRCULAR QUEUE OF API KEYS
                                                                      SET raw summary = make gemini request(prompt, api key)
                                                                      RETURN format summary(clean summary(raw summary),
FUNCTION clean summary(summary):
                                                                summary length)
  REPLACE markdown, HTML, chatty phrases WITH ""
                                                                    EXCEPT:
  NORMALIZE bullet points, whitespace
                                                                      ROTATE USED KEYS
  RETURN TRIM(summary)
                                                                      IF LAST KEY: RAISE "All keys failed"
FUNCTION make gemini request(prompt, api key) WITH
                                                                FUNCTION ask question(question, top chunks):
RETRY (3 ATTEMPTS):
                                                                  IF question CONTAINS ["title", "author", "journal", "date",
  SET url =
"https://generativelanguage.googleapis.com/v1/models/{MODE
                                                                     SET prompt = "Answer in 50 words:\nQuestion:
L}:generateContent?key={api key}"
                                                                {question}\nExcerpts:\n" + JOIN(top chunks)
  SET payload = {"contents": [{"parts": [{"text": "Expert
                                                                  ELSE:
                                                                     SET prompt = "Answer in 100 words with bullet
summarizer. {prompt}"}]}]
                                                                points:\nQuestion: {question}\nExcerpts:\n" + JOIN(top_chunks)
  SEND POST request TO url WITH payload
  IF success:
                                                                  FOR EACH api key IN USED KEYS:
    RFTURN
response["candidates"][0]["content"]["parts"][0]["text"]
                                                                       SET raw_answer = make_gemini_request(prompt, api_key)
  ELSE:
                                                                       RETURN format summary(clean summary(raw answer))
    RAISE error
                                                                     EXCEPT:
                                                                      ROTATE USED KEYS
FUNCTION summarize with Ilm(chunks,
                                                                      IF LAST KEY: RAISE "All keys failed"
summary length="medium"):
                                                                      WAIT 1 second
  IF summary length IS "short":
    SET prompt = "Summarize in 1-2 sentences:\n" +
                                                                FUNCTION format summary(summary, summary length="medium"):
JOIN(chunks[:51)
                                                                  IF summary length IN ["medium", "long"]:
  ELSE IF summary length IS "medium":
                                                                    RETURN TRIM(summary)
    SET prompt = "Summarize in 2-3 paragraphs:\n" +
                                                                  SET words = SPLIT(summary)
JOIN(chunks[:10])
                                                                  IF LEN(words) > 100:
                                                                    SET summary = JOIN(words[:100]) + "..."
  ELSE IF summary_length IS "long":
                                                                  SET sections = SPLIT(summary ON headings)
    SET prompt = "Summarize in 6-8 paragraphs with
                                                                  SET formatted = ""
headings:\n" + JOIN(chunks[:30])
                                                                  FOR EACH section IN sections:
  ELSE:
                                                                     IF MATCH "kev: content":
    SET prompt = "Summarize in 2-3 paragraphs:\n" +
                                                                      APPEND "{kev}:\n{content}\n" TO formatted WITH bullet
JOIN(chunks[:5])
                                                                points
                                                                    ELSE:
                                                                      APPEND section TO formatted
                                                                  RETURN TRIM(formatted)
```

```
# Import necessary libraries
import PDF processing library
import vector libraries (numpy, faiss)
import embedding model library
import components from Ilm utils
# Constants
CHUNK SIZE = 150 # words
CHUNK OVERLAP = 25 # words
embedder = load SentenceTransformer model
# Document Processing Functions
function chunk text(text):
  Split text into words
  Create overlapping chunks of CHUNK SIZE words with CHUNK OVERLAP
  Return list of chunks
function process pdf(pdf path):
  Open PDF document
  Extract full text from all pages
  Split text into chunks using chunk text()
  Generate embeddings for all chunks
  Create FAISS vector index
  Add embeddings to index
  Return chunks and index
# Retrieval Functions
function get top chunks(query, chunks, index, top k=5):
  Encode query using the embedder
  Perform vector search with FAISS index
  Return top k most similar chunks
# RAG Implementation
function answer_with_rag(query, chunks, index, top_k=5):
  Get top k relevant chunks using get top chunks()
  Log debug info
  For each API key in rotation:
       Get answer using ask question() from llm utils
       Log success
       Return response
     Catch exceptions:
       Log failure
       Rotate to next API key
```

If all keys failed:

Raise exception

Wait briefly before retrying

rag utils.pv



Import necessary libraries import FastAPI and related components import MongoDB client main.py import PDF processing libraries import custom modules (rag_utils, llm_utils) import file handling and logging # Configure application setup FastAPI app configure logging setup PDF generation tool configure CORS middleware # Constants and Configuration UPLOAD DIR = "uploads" ensure directory exists # Setup MongoDB connection connect to MongoDB using connection string setup database collections for papers, summaries, and chat history # API Endpoints @app.post("/upload/") function upload pdf(file): Save uploaded file to UPLOAD DIR Process PDF using process pdf() from rag utils Generate initial summary using summarize with Ilm() Create unique paper id Save FAISS index to disk Store paper metadata in papers collection Store summary in summaries collection Return paper id, filename, summary and date @app.post("/summarize/") function summarize pdf(file, summary length): Save uploaded file to UPLOAD DIR Process PDF using process pdf() Generate summary of specified length using summarize with Ilm() Create unique paper_id Save FAISS index to disk Store paper metadata in papers_collection Store summary with length info in summaries collection Return paper id and summary @app.get("/summaries/") function get summaries(limit=10): Retrieve most recent papers from papers collection Initialize empty result list and tracking set For each paper: Skip if already seen Get latest summary for this paper Format summary data with metadata Get chat history for this paper Add to result list

Return result list

@app.post("/chat/") function chat with paper(question, paper id): Validate inputs Find paper in database Get or reload chunks and index Get top relevant chunks using get_top_chunks() Generate answer using ask question() Store chat entry in chat history collection Return answer @app.get("/history/{paper_id}") function get paper history(paper id): Validate paper id Find paper in database Get summary for paper Get chat history for paper Format response with paper metadata, summary and chat history Return formatted response @app.post("/download-summary/") function download summary(request): Parse request to get summary text

@app.delete("/delete-summary/{summary_id}") function delete_summary(summary_id):
Find summary in database
Delete related chat history
Delete summary
Return success message

Generate HTML content with styling Set PDF generation options

Generate PDF file using pdfkit

Return file download response

@app.post("/api/define")
function define_word(request):
 Get word from request
 Check spelling and get correction
 Query dictionary API
 Parse response for definitions
 Return formatted definitions or error message



Achievements

- •Developed :Delivered a working version of the application that demonstrates core features, user flow.
- •Successfully handled summarization of research Paper across domains

Achieved reliable and context-aware summarization for a wide range of topics including computer science, engineering, and social sciences.

- Integrated helpful utilities .
- Helps researchers, students, and professionals quickly understand key insights, saving time and boosting efficiency.
- Personalized User Experience:, history tracking, saved summaries, and customization options like summary length and voice mode.
- Enhanced Accessibility: Voice input and text-to-speech features improve usability for differently-abled users and those preferring audio output.
- Multilingual Document Summarization : Allows uploading papers in any language (e.g., Telugu) and generates summaries in English, enabling cross-language understanding of academic content.
- ❖ Interactive Learning Environment : Al Chat Board supports conversation-based queries, deeper insights, definitions, and explanation of paper content.
- User Feedback Integration: Feedback system helps gather user input to improve system quality and user satisfaction over time.

Key Challenges

Preserving technical accuracy

Summarizing complex academic language without losing key details.

Speed optimization for large PDFs and documents

Ensuring quick responses and fast loading, even with documents that have a lot of pages.

Designing an intuitive yet powerful UI

Designing a simple, easy-to-use interface while including advanced features like voice-to-text from AI chat bot, and History saving for all users.

CONCLUSION

A fully functional, web-based application has been developed to summarize research papers efficiently. The platform leverages **Large Language Models (LLMs)** and **Retrieval-Augmented Generation (RAG)** techniques in the backend to generate high-quality, context-aware summaries. The frontend is built using **React**, providing a clean, responsive, and user-friendly interface. The system supports various features like adjustable summary length, dark/light mode, an integrated dictionary, chatbot assistance, and user history tracking — making it a powerful tool for students, researchers, and professionals alike.

It efficiently summarizes research papers and provides a user-friendly interface with rich academic support features.

New Skills Learned:

- •Implementation of LLMs and Retrieval-Augmented Generation.
- •UX design tailored for research-heavy tools.
- •Working with frontend-backend integration and secure authentication.
- •Managing user data securely with database systems.



Thank you!