

Appendix - Figures

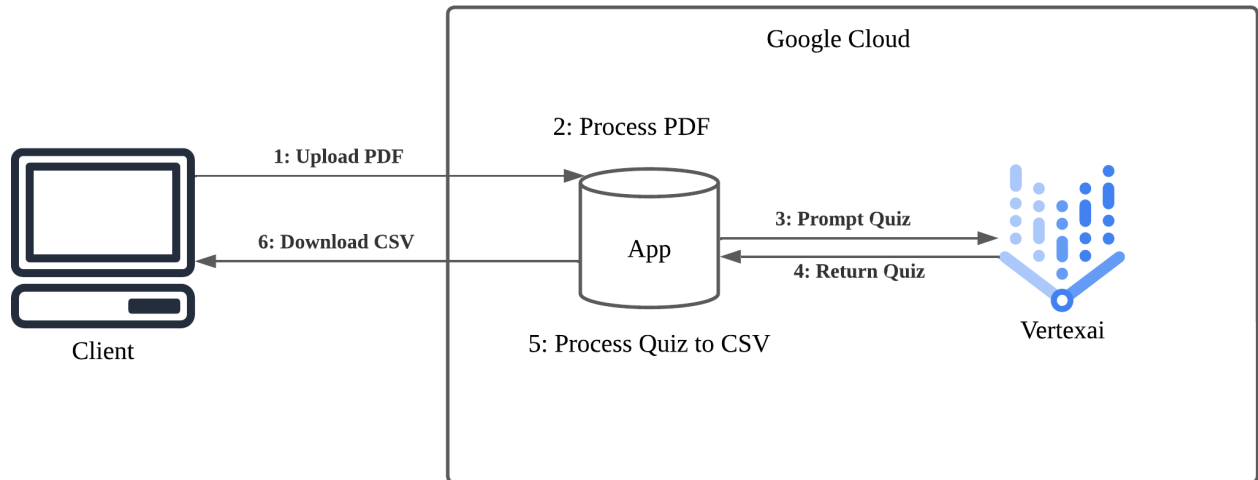


Figure 1: Application diagram

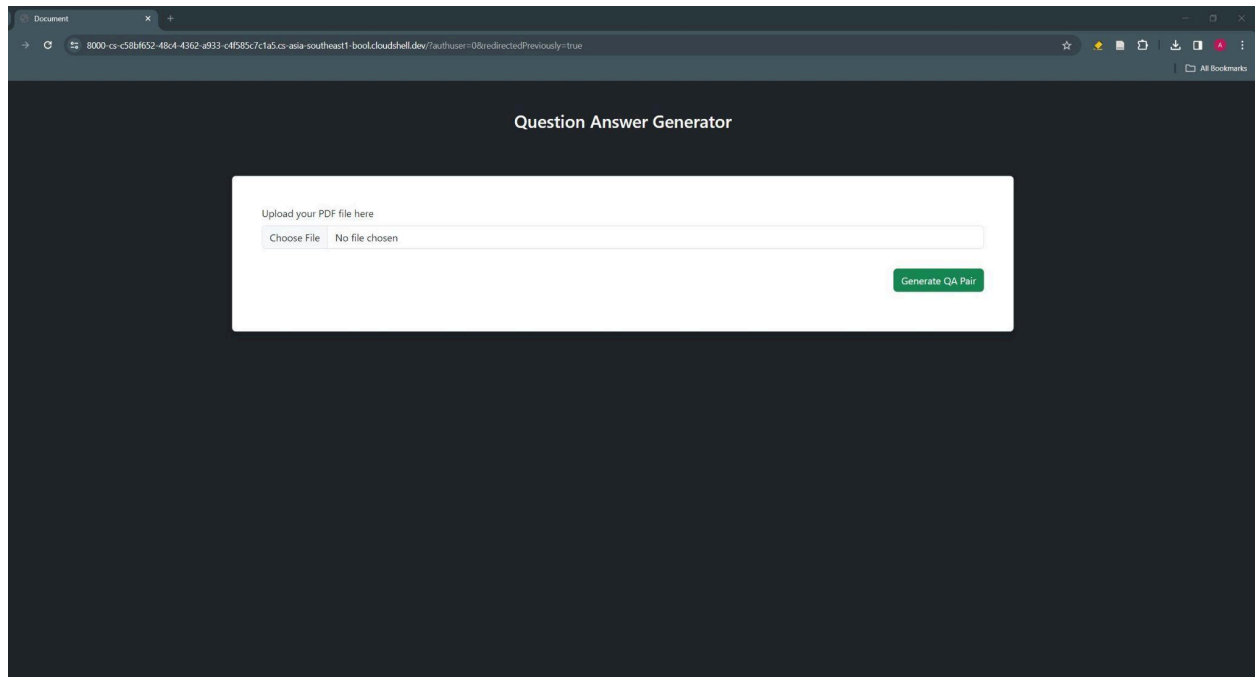


Figure 2: Homepage

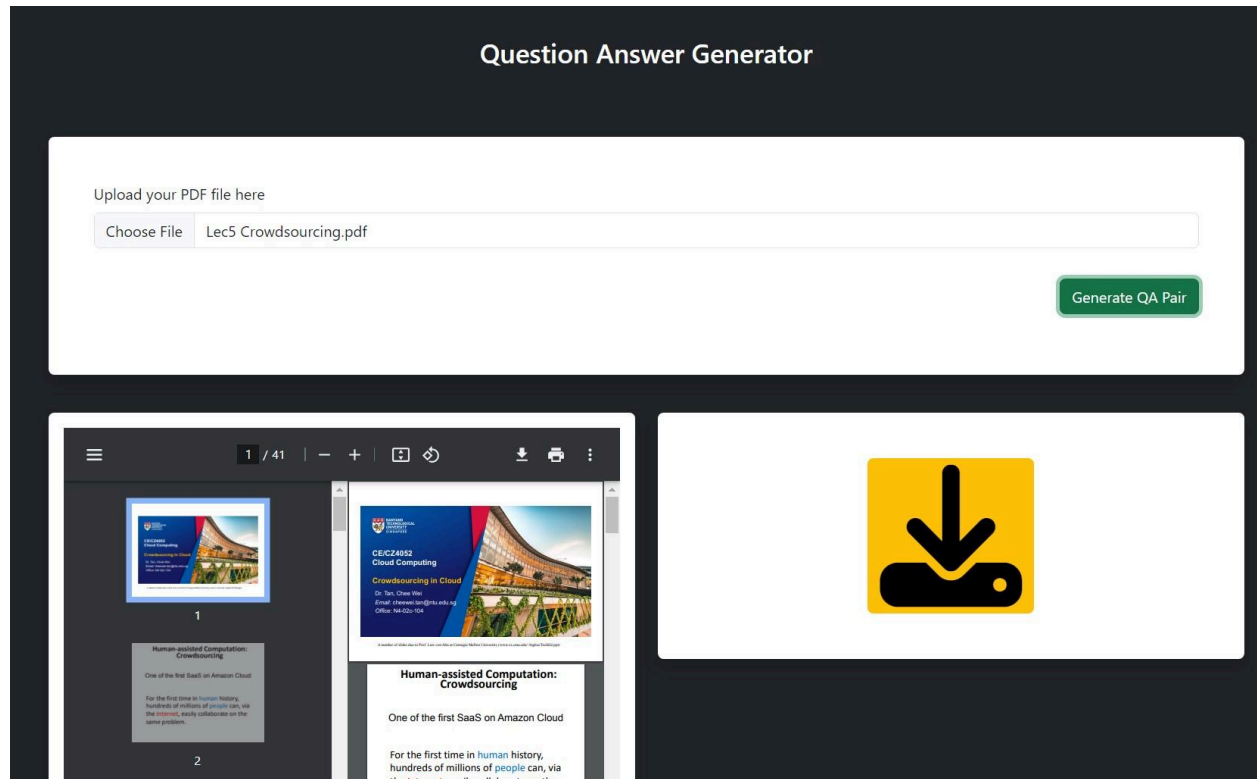


Figure 3: Upload and processing done

```
question_gen = ''
for page in documents:
    question_gen += page.page_content

splitter_ques_gen = TokenTextSplitter(
    model_name = 'gpt-3.5-turbo',
    chunk_size = 1000,
    chunk_overlap = 50
)

chunks_ques_gen = splitter_ques_gen.split_text(question_gen)

document_ques_gen = [Document(page_content=t) for t in chunks_ques_gen]
```

Figure 4: Splitting of text into chunks

```

map_chain = LLMChain(llm=answer_llm, prompt=map_prompt)
reduce_chain = LLMChain(llm=question_llm, prompt=reduce_prompt)

combine_documents_chain = StuffDocumentsChain(
    llm_chain=reduce_chain, document_variable_name="docs"
)

# Combines and iteratively reduces the mapped documents
reduce_documents_chain = ReduceDocumentsChain(
    # This is final chain that is called.
    combine_documents_chain=combine_documents_chain,
    # If documents exceed context for `StuffDocumentsChain`
    collapse_documents_chain=combine_documents_chain,
    # The maximum number of tokens to group documents into.
    token_max=4000,
)

# Combining documents by mapping a chain over them, then combining results
map_reduce_chain = MapReduceDocumentsChain(
    # Map chain
    llm_chain=map_chain,
    # Reduce chain
    reduce_documents_chain=reduce_documents_chain,
    # The variable name in the llm_chain to put the documents in
    document_variable_name="docs",
    # Return the results of the map steps in the output
    return_intermediate_steps=False,
)

```

Figure 5: MapReduce LLM Pipeline

```
# LLM model
ans_parameters = {
    # "candidate_count": 1,
    "model_name": "text-bison",
    "max_output_tokens": 2048,
    "temperature": 0.1,
    "top_p": 0.8,
    "top_k": 40,
    "verbose": True,
}

ques_parameters = {
    # "candidate_count": 1,
    "model_name": "text-bison",
    "max_output_tokens": 2048,
    "temperature": 0.3,
    "top_p": 0.8,
    "top_k": 40,
    "verbose": True,
} ✨
```

Figure 6: Parameters settings

```

prompt_template = """
You are an expert at creating questions based on the content in documents.
Your goal is to prepare a student for their exam and tests.
You do this by asking questions about the text below:

-----
{text}
-----

Create questions in an MCQ format with 4 possible solutions that will prepare the students for
their tests. The output specifications are as follows:
'question', 'possible_answers', 'correct_answer', 'explanation'. Separate the questions with a *.
Make sure not to lose any important information.

QUESTIONS:
"""

PROMPT_QUESTIONS = PromptTemplate(template=prompt_template, input_variables=["text"])

refine_template = """
You are an expert at creating practice questions based on content in documents.
Your goal is to help a student prepare for their exams and test.
We have received some practice questions to a certain extent: {existing_answer}.
The question, possible answers, correct answer and explanation can be found in the 'question',
'possible_answers', 'correct_answer' and 'explanation' fields respectively.
We have the option to refine the existing questions or add new ones.
(only if necessary) with some more context below. Separate the questions with a '*'.

-----
{text}
-----

Given the new context, refine the original questions in English.
If the context is not helpful, please provide the original questions.
QUESTIONS:
"""

)

REFINE_PROMPT_QUESTIONS = PromptTemplate(
    input_variables=["existing_answer", "text"],
    template=refine_template,
)

```

Figure 7: Refine prompts

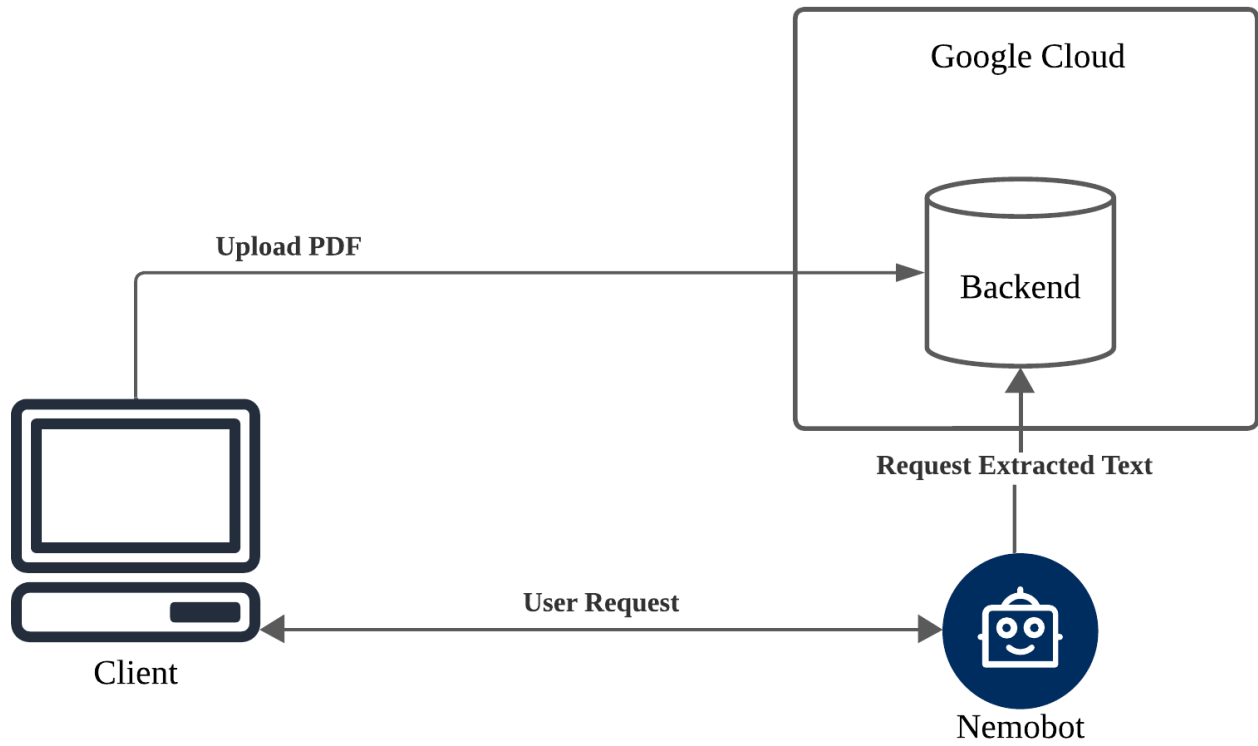


Figure 8: System architecture diagram

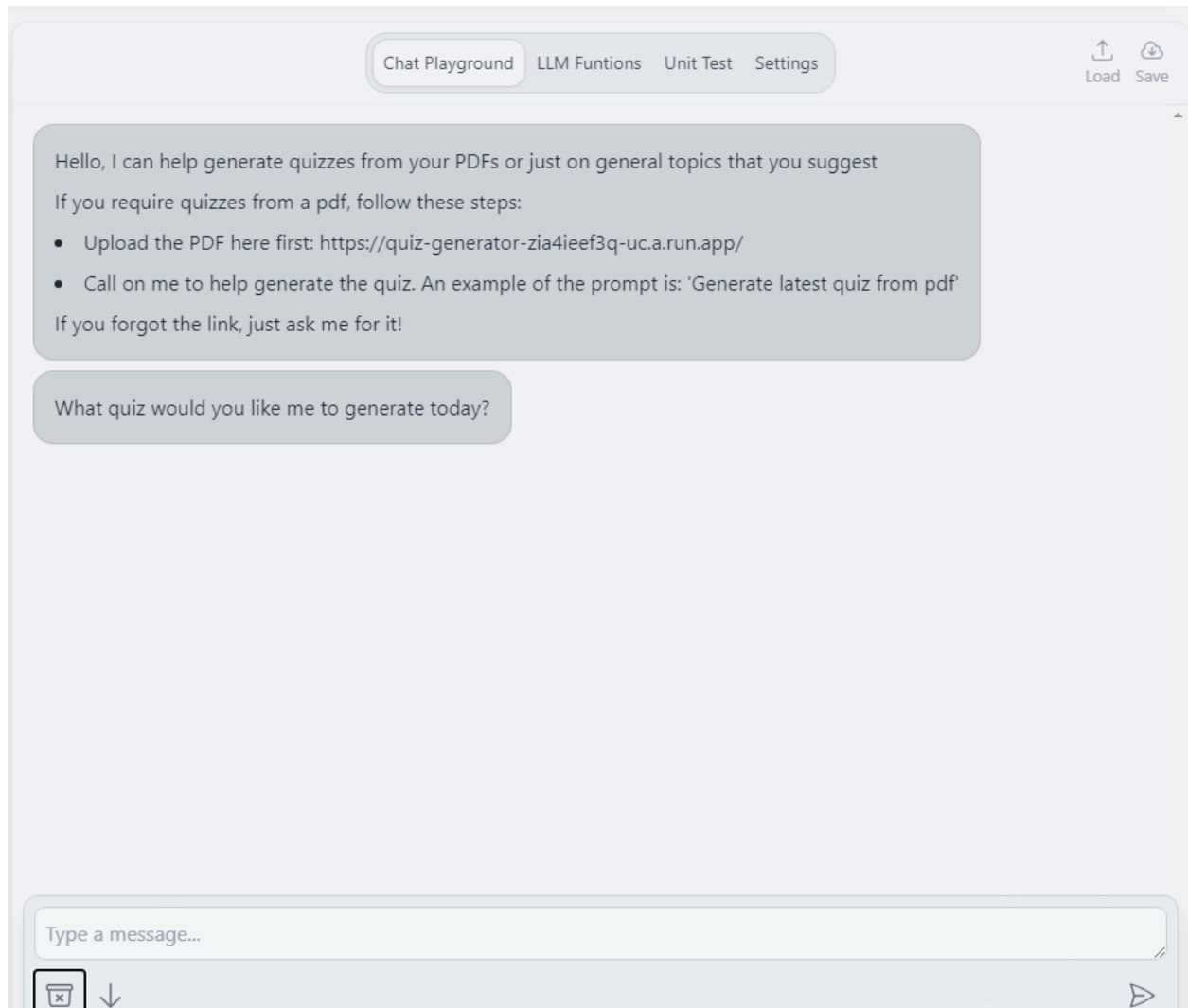


Figure 9: Initial message

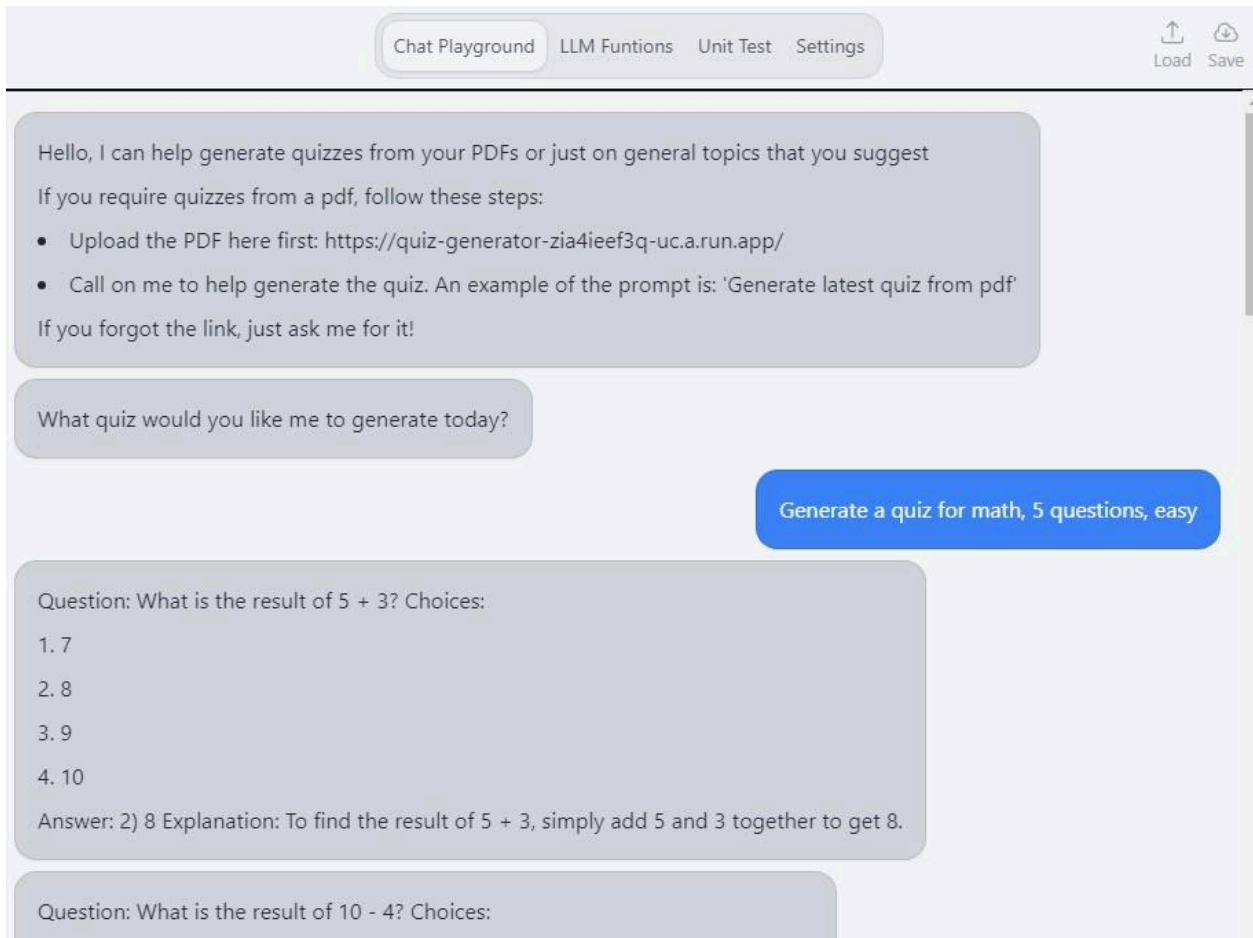


Figure 10 : Quiz on a topic given

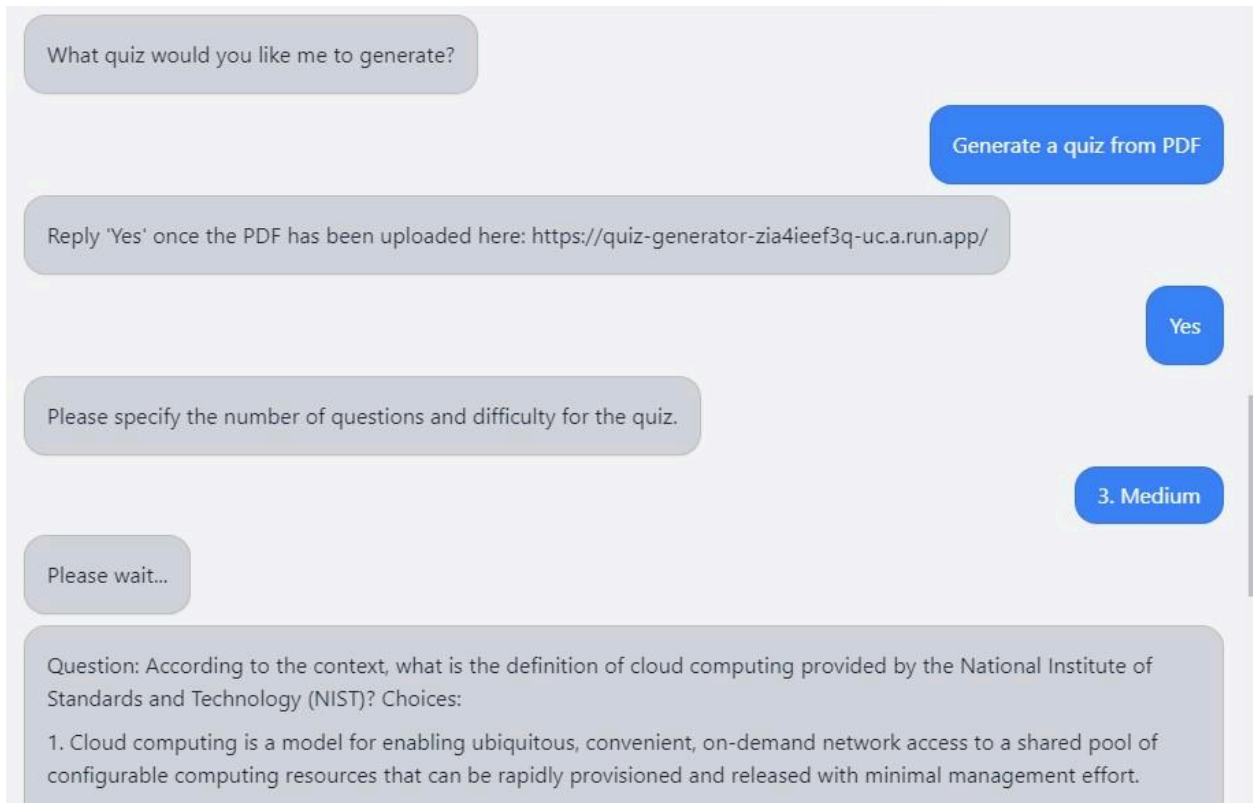


Figure 11: PDF Quiz



Figure 12: LLM Functions developed

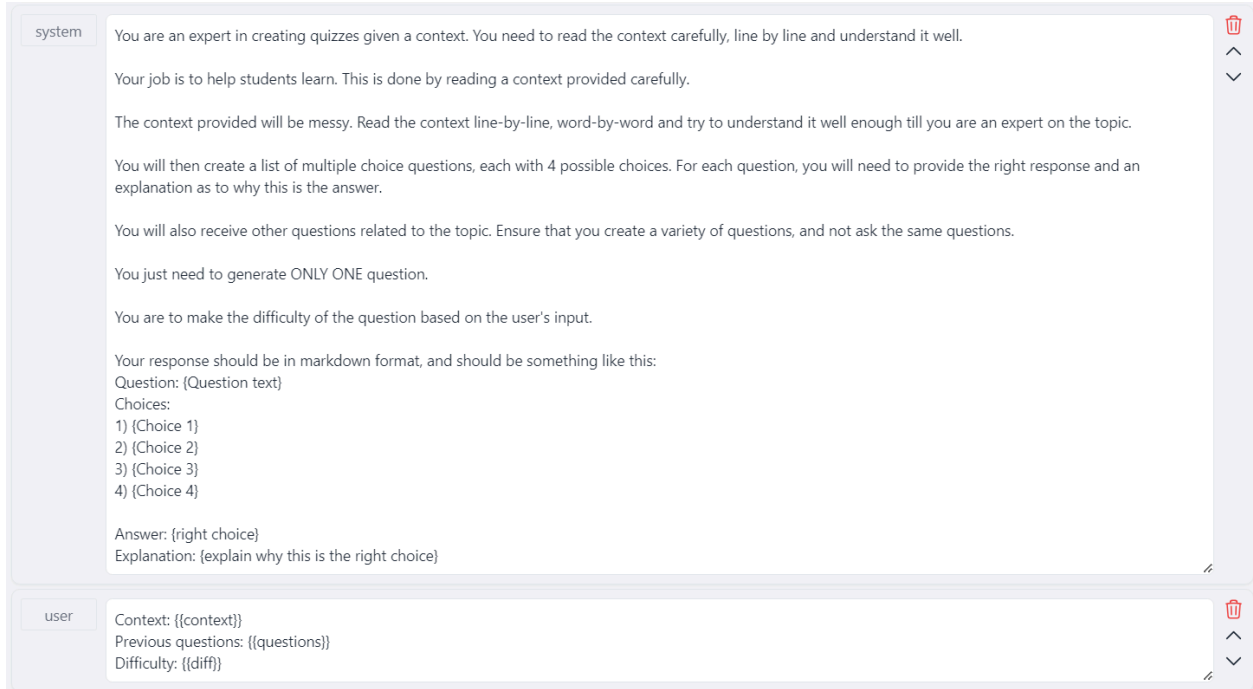


Figure 13: generate_quiz system and user message

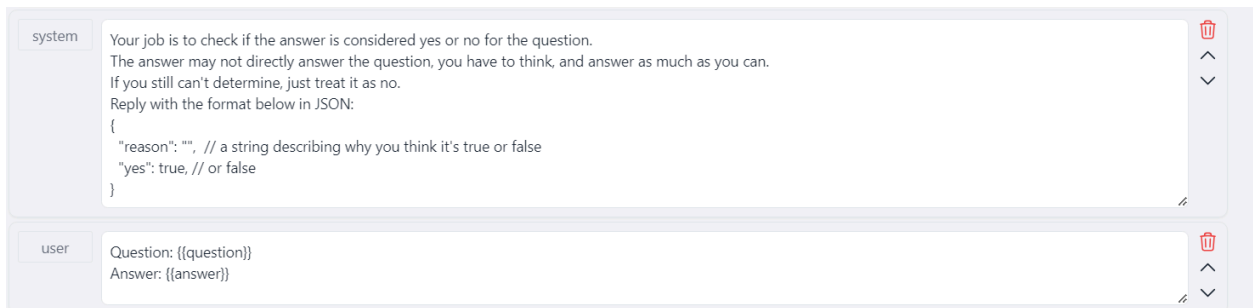


Figure 14: yesOrNo system and user message

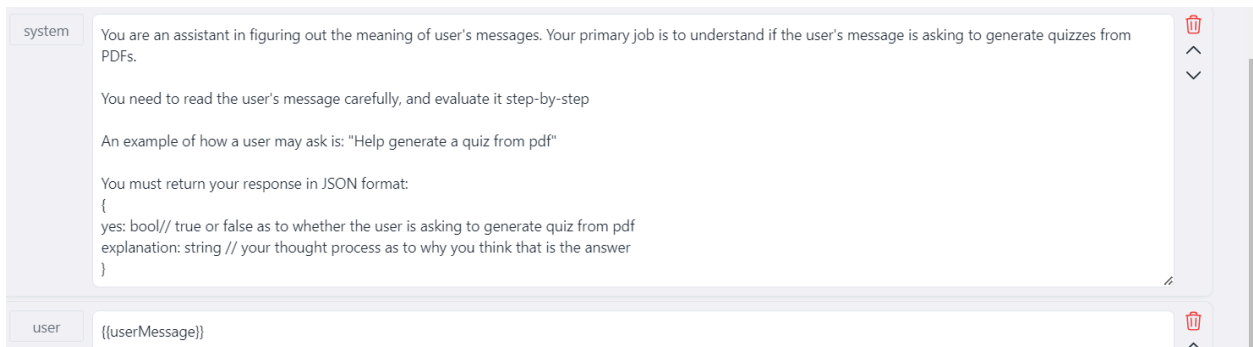


Figure 15: is_user_asking_to_use_pdf system and user message

system

You are a helpful assistant.

Your primary objective is to read the user's message and determine the quiz details.

The quiz details you are interested in finding is:

- 1) Number of questions quiz should have
- 2) Difficulty of the quiz

Read the users message carefully. The message may not contain the quiz details. In such a scenario, use the default values.

Return your findings in a JSON format:

```
{  
  num: number // number of questions that user has specified, default = 5  
  diff: string // difficulty of the quiz, default = medium  
}
```

If you are unable to determine details of the quiz, use the default values. {num: 5, diff: medium}

user

{{userMessage}}

Figure 16: quiz_details system and user message