GPT4All is described as a free-to-use, locally running, privacy-aware chatbot. It is an ecosystem that allows users to train and deploy large language models on consumer-grade CPUs without the need for a GPU or internet connection. The GPT4All model is a 4GB file that can be downloaded and integrated into the open-source GPT4All ecosystem software provided by Nomic Al.

The workflow of using GPT4All involves several steps. First, the GPT4All model is loaded. Then, Langchain is used to retrieve and load the relevant documents. These documents are then split into smaller chunks that can be processed by embeddings. Embeddings are numerical representations of information that capture the semantic meaning of the embedded content.

Next, FAISS is utilized to create a vector database with the embeddings. This vector database enables semantic search, allowing users to perform similarity searches based on the question they want to ask GPT4AII. The search result serves as context for the question.

Finally, the question and context are fed to GPT4All using Langchain, and the system waits for the answer. The answer is generated by the GPT4All model based on the provided question and context.

In this project, the emphasis is on using lightweight embeddings rather than heavy GPU models, making it feasible to run the system on consumer-grade CPUs. The Alpaca native model and LlamaCppEmbeddings from Langchain are specifically used for this purpose. The entire process is explained in a step-by-step manner for easy implementation.

Coding steps (windows)

1.Create a new folder

```
mkdir GPT4ALL_Fabio cd GPT4ALL Fabio
```

2. Set up the virtual environment so that it won't affect the other functions

```
python -m venv .venv
```

3.activate it venv\Scripts\activate

4.Download the different library

```
pip install pygpt4all==1.0.1
pip install pyllamacpp==1.0.6
pip install langchain==0.0.149
pip install unstructured==0.6.5
pip install pdf2image==1.16.3
pip install pytesseract==0.3.10
```

```
pip install pypdf==3.8.1
pip install faiss-cpu==1.7.4
```

5. Download ready to use GPT4ALL model

https://huggingface.co/mrgaang/aira/blob/main/gpt4all-converted.bin

We also need the model for embedding, this model can run on CPU Without crashing https://huggingface.co/Pi3141/alpaca-native-7B-ggml/tree/397e872bf4c83f4c642317a5bf65ce84a105786e

- 6.Go here https://github.com/abetlen/llama-cpp-python/releases and look for the complied wheel for your architecture and python version — you MUST take Wheels Version 0.1.49
- 7. Then i put the two model file which we downloaded earlier, in model file in lib folder in the folder created in first step
- 8.Create a new python file an import GPT4ALL model in it, do this in the same folder created earlier. from pygpt4all.models.gpt4all import GPT4All.

9. This is the python binding for our model. Now we can call it and start asking. Let's try a creative one. We create a function that read the callback from the model, and we ask GPT4All to complete our sentence.

```
def new_text_callback(text):
    print(text, end="")

model = GPT4All('./models/gpt4all-converted.bin')
model.generate("Once upon a time, ", n_predict=55,
    new_text_callback=new_text_callback)
```

Run the code by python pygpt4all test.py

The first statement is telling our program where to find the model (remember what we did in the section above) The second statement is asking the model to generate a response and to complete our prompt "Once upon a time, ".

10. Create a new python file , namely my_langchain.py. And run the following code in it.

```
# Import of langchain Prompt Template and Chain
from langchain import PromptTemplate, LLMChain
# Import llm to be able to interact with GPT4All directly from langchain
from langchain.llms import GPT4All
# Callbacks manager is required for the response handling
from langchain.callbacks.base import CallbackManager
from langchain.callbacks.streaming stdout import
StreamingStdOutCallbackHandler
local path = './models/gpt4all-converted.bin'
callback manager = CallbackManager([StreamingStdOutCallbackHandler()])
Now we only need to link together our template, the question and the language model.
template = """Question: {question}
Answer: Let's think step by step on it.
prompt = PromptTemplate(template=template, input variables=["question"])
# initialize the GPT4All instance
llm = GPT4All(model=local path, callback manager=callback_manager,
verbose=True)
# link the language model with our prompt template
llm chain = LLMChain(prompt=prompt, llm=llm)
```

```
# Hardcoded question
question = "What Formula 1 pilot won the championship in the year Leonardo
di Caprio was born?"

# User imput question...
# question = input("Enter your question: ")

#Run the query and get the results
llm chain.run(question)
```

Then run it by python my langchain.py

11. Answering questions related to document using langchain and gpt4all Create a new file name my_knowledge_qna.py and run the following code

```
from langchain import PromptTemplate, LLMChain
from langchain.llms import GPT4All
from langchain.callbacks.base import CallbackManager
from langchain.callbacks.streaming_stdout import
StreamingStdOutCallbackHandler
# function for loading only TXT files
from langchain.document_loaders import TextLoader
# text splitter for create chunks
from langchain.text_splitter import RecursiveCharacterTextSplitter
# to be able to load the pdf files
from langchain.document_loaders import UnstructuredPDFLoader
from langchain.document_loaders import PyPDFLoader
from langchain.document_loaders import DirectoryLoader
```

```
# Vector Store Index to create our database about our knowledge
from langchain.indexes import VectorstoreIndexCreator
# LLamaCpp embeddings from the Alpaca model
from langchain.embeddings import LlamaCppEmbeddings
# FAISS library for similaarity search
from langchain.vectorstores.faiss import FAISS
import os #for interaaction with the files
import datetime
# assign the path for the 2 models GPT4All and Alpaca for the embeddings
gpt4all path = './models/gpt4all-converted.bin'
llama path = './models/ggml-model-q4 0.bin'
# Calback manager for handling the calls with the model
callback manager = CallbackManager([StreamingStdOutCallbackHandler()])
# create the embedding object
embeddings = LlamaCppEmbeddings(model path=llama path)
# create the GPT4All llm object
llm = GPT4All(model=gpt4all path, callback manager=callback manager,
verbose=True)
# Split text
def split chunks(sources):
chunks = []
splitter = RecursiveCharacterTextSplitter(chunk size=256,
chunk overlap=32)
for chunk in splitter.split documents(sources):
chunks.append(chunk)
return chunks
def create index(chunks):
texts = [doc.page content for doc in chunks]
metadatas = [doc.metadata for doc in chunks]
search index = FAISS.from texts(texts, embeddings, metadatas=metadatas)
return search index
def similarity search(query, index):
# k is the number of similarity searched that matches the query
# default is 4
matched docs = index.similarity search(query, k=3)
sources = []
for doc in matched docs:
sources.append(
"page_content": doc.page_content,
```

```
"metadata": doc.metadata,
}
)
return matched docs, sources
# get the list of pdf files from the docs directory into a list format
pdf folder path = './docs'
doc list = [s for s in os.listdir(pdf folder path) if s.endswith('.pdf')]
num of docs = len(doc list)
# create a loader for the PDFs from the path
general start = datetime.datetime.now() #not used now but useful
print("starting the loop...")
loop start = datetime.datetime.now() #not used now but useful
print("generating fist vector database and then iterate with .merge from")
loader = PyPDFLoader(os.path.join(pdf folder path, doc list[0]))
docs = loader.load()
chunks = split chunks(docs)
db0 = create index(chunks)
print("Main Vector database created. Start iteration and merging...")
for i in range(1, num of docs):
print(doc list[i])
print(f"loop position {i}")
loader = PyPDFLoader(os.path.join(pdf folder path, doc list[i]))
start = datetime.datetime.now() #not used now but useful
docs = loader.load()
chunks = split chunks(docs)
dbi = create index(chunks)
print("start merging with db0...")
db0.merge from(dbi)
end = datetime.datetime.now() #not used now but useful
elapsed = end - start #not used now but useful
#total time
print(f"completed in {elapsed}")
print("----")
loop end = datetime.datetime.now() #not used now but useful
loop elapsed = loop end - loop start #not used now but useful
print(f"All documents processed in {loop elapsed}")
print(f"the daatabase is done with {num of docs} subset of db index")
print("----")
print(f"Merging completed")
print("----")
print("Saving Merged Database Locally")
# Save the databasae locally
db0.save local("my faiss index")
print("----")
print("merged database saved as my faiss index")
general end = datetime.datetime.now() #not used now but useful
```

general_elapsed = general_end - general_start #not used now but useful
print(f"All indexing completed in {general_elapsed}")
print("----")

```
Please use the following context to answer questions.

Contacts mostforce by 2025 -spacially those underrepresented in the field. And we 're helping an additional 24 countries with substantial cyberse-equity workforce shortages close their gaps too.

Viva, we're building an employee experience platform that brings together communications, knowledge, learning, resources, and insights in the flow of work to empower employees and strengthen their connection to their company's mission and culture.

Hub provides detailed reports on our environmental data, our political activities, our workforce deengarphics, our human rights work, and more.

Tuestion: In 2022, our employees gave how much to over 22,000 nanprofits?

Answers: let's think step by step. Firstly, we need some information about the context given in order to answer this question accurately: We are trying to build an employee experience platform that will strengthen connection of their company's mission and culture with employees which should give us enough clues on what our current morthforce looks like (gender/race etc.)

With this, we can infer some details about the political activities. However, more investigation is needed for environmental data or human rights' works as they require a lot of effort to gather information from various sources such as social media posts and corporate reports which are not mentioned in our cont ext provided above. [end of text]

Llama_print_timings: such with the such as the such
```

12.Put the following code inside a python file *db_loading.py* and run it with the command from terminal *python3 db_loading.py*

```
from langchain import PromptTemplate, LLMChain
from langchain.llms import GPT4All
from langchain.callbacks.base import CallbackManager
from langchain.callbacks.streaming stdout import
StreamingStdOutCallbackHandler
# function for loading only TXT files
from langchain.document loaders import TextLoader
# text splitter for create chunks
from langchain.text splitter import RecursiveCharacterTextSplitter
# to be able to load the pdf files
from langchain.document loaders import UnstructuredPDFLoader
from langchain.document loaders import PyPDFLoader
from langchain.document loaders import DirectoryLoader
# Vector Store Index to create our database about our knowledge
from langchain.indexes import VectorstoreIndexCreator
# LLamaCpp embeddings from the Alpaca model
from langchain.embeddings import LlamaCppEmbeddings
# FAISS library for similaarity search
from langchain.vectorstores.faiss import FAISS
import os #for interaaction with the files
import datetime
```

```
# TEST FOR SIMILARITY SEARCH
# assign the path for the 2 models GPT4All and Alpaca for the embeddings
gpt4all path = './models/gpt4all-converted.bin'
llama path = './models/ggml-model-q4 0.bin'
# Calback manager for handling the calls with the model
callback manager = CallbackManager([StreamingStdOutCallbackHandler()])
# create the embedding object
embeddings = LlamaCppEmbeddings(model path=llama path)
# create the GPT4All llm object
11m = GPT4All(model=gpt4all path, callback manager=callback manager,
verbose=True)
# Split text
def split chunks(sources):
chunks = []
splitter = RecursiveCharacterTextSplitter(chunk size=256,
chunk overlap=32)
for chunk in splitter.split documents(sources):
chunks.append(chunk)
return chunks
def create index(chunks):
texts = [doc.page content for doc in chunks]
metadatas = [doc.metadata for doc in chunks]
search index = FAISS.from texts(texts, embeddings, metadatas=metadatas)
return search index
def similarity search(query, index):
# k is the number of similarity searched that matches the query
# default is 4
matched\ docs = index.similarity\ search(query,\ k=3)
sources = []
for doc in matched docs:
sources.append(
"page content": doc.page content,
"metadata": doc.metadata,
```

```
# Load our local index vector db
index = FAISS.load_local("my_faiss_index", embeddings)
# Hardcoded question
query = "In 2022, our employes donates how much to 32,000 nonprofit
organisation"
docs = index.similarity_search(query)
# Get the matches best 3 results - defined in the function k=3
print(f"The question is: {query}")
print("Here the result of the semantic search on the index, without
GPT4All..")
print(docs[0])
```

Run the following code to get the answer

```
# Load our local index vector db
```

```
index = FAISS.load local("my faiss index", embeddings)
```

```
# create the prompt template
template = """
Please use the following context to answer questions.
Context: {context}
Question: {question}
Answer: Let's think step by step."""
# Hardcoded question
question = "What is a PLC and what is the difference with a PC"
matched_docs, sources = similarity_search(question, index)
# Creating the context
context = "\n".join([doc.page_content for doc in matched_docs])
```

```
# instantiating the prompt template and the GPT4All chain
```

```
prompt = PromptTemplate(template=template, input_variables=["context",
    "question"]).partial(context=context)
```

```
llm chain = LLMChain(prompt=prompt, llm=llm)
```

Print the result

```
print(llm chain.run(question))
```

Then run the file



