# policy\_chatbot

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### 0.0.1 Nestle Policy Chatbot - Simplifearn Submission

Name: Shrawanika WakdeEmail: shrawanika@gmail.com

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## 1 Project

**Task:** Your task is to develop a conversational chatbot. This chatbot must answer queries about Nestlé's HR reports efficiently. Use Python libraries, OpenAI's GPT model, and Gradio UI. These tools will help you create a user-friendly interface. This interface will extract and process information from documents. It will provide accurate responses to user queries.

Action: • Import essential tools and set up OpenAI's API environment. • Load Nestle's HR policy using PyPDFLoader and split it for easy processing. • Create vector representations for text chunks using Chroma dB and OpenAI's embeddings. • Build a question-answering system using the GPT-3.5 Turbo model to retrieve answers from text chunks. • Create a prompt template to guide the chatbot in understanding and responding to users. • Use Gradio to build a user-friendly chatbot interface, enabling interaction and information retrieval.

#### 1.1 Challenges Encountered

#### 1. Using static file input instead of UI upload

- Currently, the system relies on static PDF files rather than allowing users to upload documents through the UI.
- Because of Gradio instability in JupyterLab, not using Gradio Upload Button which uses Blocks
- Using Gradio Blocks sometimes causes random errors, making the interface unreliable in the lab environment.

#### 2. Using FAISS instead of Chroma:

- ChromaDB requires SQLite, but the lab environment has an outdated, unsupported SQLite version.
- This prevents creation of both in-memory and persistent vector stores.
- Upgrading SQLite requires admin access, which is unavailable in the lab.
- Attempts to use duckdb+parquet as a workaround fail because the installed Chroma version still depends on SQLite internally.

- Due to SQLite constraints, FAISS is used as an alternative in-memory vector store.
- FAISS avoids system-level dependencies, allowing embeddings and retrieval to work reliably in the lab environment.

```
[21]: # 1. Imports
      from langchain.document_loaders import PyPDFLoader
      from langchain.text_splitter import RecursiveCharacterTextSplitter
      from langchain.embeddings.openai import OpenAIEmbeddings
      from langchain.vectorstores import FAISS
      from langchain.prompts import PromptTemplate
      from langchain.chat_models import ChatOpenAI
      from langchain.chains import RetrievalQA
      # 2. Load PDF
      loader = PyPDFLoader("inputs/Nestle HR Policy.pdf") # replace with your PDFL
       \hookrightarrow path
      documents = loader.load()
      # 3. Split Text
      text_splitter = RecursiveCharacterTextSplitter(chunk_size=1000,__
      ⇔chunk_overlap=100)
      docs = text_splitter.split_documents(documents)
      # 4. Create embeddings and inmemory vector DB
      embeddings = OpenAIEmbeddings()
      vectordb = FAISS.from_documents(docs, embeddings)
      # 5. Prompt template
      prompt = PromptTemplate(
          input_variables=["context", "question"],
          template="""You are a helpful HR assistant. Use the context below to answer
       Context: {context}\n\nQuestion: {question}\nAnswer:"""
      # 6. Retriever Chain
      retriever = vectordb.as_retriever(search_kwargs={"k": 3})
      qa = RetrievalQA.from_chain_type(
          llm=ChatOpenAI(model_name="gpt-3.5-turbo", temperature=0),
          chain_type="stuff",
          retriever=retriever,
          chain_type_kwargs={"prompt": prompt}
```

```
[22]: # Gradio Interface
import gradio as gr
print(gr.__version__)
```

```
def answer_question(query):
          return qa.run(query)
      iface = gr.Interface(
          fn = answer_question,
          title = "Nestle HR Policy Chatbot",
          inputs = gr.Textbox(label="Ask about HR Policy"),
          outputs = gr.Textbox(label="Bot Answer"),
          flagging_options = ["incorrect response"],
          flagging_dir = "reported_responses" # Reported repsonses will be saved_
       →under this folder in log.csv
      iface.launch(share=True) # Without share=true, it doesn not load interface_
       →inside the Lab
     4.44.1
     Running on local URL: http://127.0.0.1:7860
     Running on public URL: https://28cc5052eeac7c985d.gradio.live
     This share link expires in 72 hours. For free permanent hosting and GPU
     upgrades, run `gradio deploy` from Terminal to deploy to Spaces
     (https://huggingface.co/spaces)
     <IPython.core.display.HTML object>
[22]:
[20]: #iface.close()
     Closing server running on port: 7860
```

[]: