### Here's the **fully working Gradio app** for a PDF Q&A chatbot using:

- Iangchain ollama for LLaMA 3 via Ollama
- langchain\_huggingface for HuggingFace embeddings
- FAISS for vector storage
- gradio for the user interface
- Correct handling of file paths (no .read() bug)

### ✓ Complete & Fixed Code (chatbot.py):-

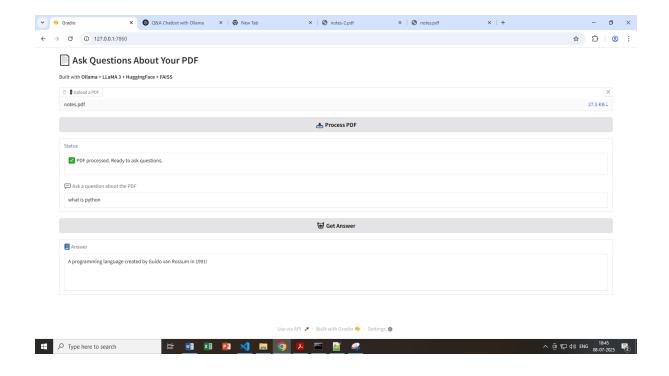
```
import gradio as gr
from langchain community.document loaders import PyPDFLoader
from langchain.text splitter import RecursiveCharacterTextSplitter
from langchain_huggingface import HuggingFaceEmbeddings
from langchain_community.vectorstores import FAISS
from langchain_ollama import OllamaLLM
from langchain.chains import RetrievalQA
# Step 1: Load PDF and split into documents
def load_pdf(path):
   loader = PyPDFLoader(path)
    docs = loader.load_and_split()
    return docs
# Step 2: Create FAISS vector store from documents
def create vectorstore(docs):
    splitter = RecursiveCharacterTextSplitter(chunk_size=500,
chunk_overlap=50)
    texts = splitter.split_documents(docs)
    embeddings = HuggingFaceEmbeddings(model_name="sentence-transformers/all-
MiniLM-L6-v2")
    vectordb = FAISS.from_documents(texts, embedding=embeddings)
    return vectordb
# Step 3: Load Ollama LLM
def load_llm():
    return OllamaLLM(model="llama3")
# Step 4: Build Retrieval QA Chain
def create_qa_chain(llm, vectordb):
    retriever = vectordb.as_retriever(search_kwargs={"k": 3})
    return RetrievalQA.from_chain_type(llm=llm, chain_type="stuff",
retriever=retriever)
# Handle PDF processing
```

```
def process_pdf(file_path):
    try:
       docs = load pdf(file path)
       vectordb = create vectorstore(docs)
        llm = load llm()
        qa_chain = create_qa_chain(llm, vectordb)
        return f" PDF processed. Ready to ask questions.", qa chain
    except Exception as e:
        return f" X Error while processing PDF: {e}", None
# Handle Q&A interaction
def answer question(question, ga chain):
    if not qa chain:
       return " \( \) Please upload and process a PDF first."
    if not question.strip():
        return "∧ Please enter a question."
    try:
        result = qa_chain.invoke({"query": question})
        return result["result"] if isinstance(result, dict) else result
    except Exception as e:
        return f" X Error while answering: {e}"
# Build Gradio UI
with gr.Blocks() as demo:
    gr.Markdown("# Ask Questions About Your PDF")
    gr.Markdown("Built with **Ollama + LLaMA 3 + HuggingFace + FAISS**")
    qa_chain_state = gr.State()
    pdf_input = gr.File(label="[ Upload a PDF", file_types=[".pdf"])
    upload_button = gr.Button(" Process PDF")
    status_output = gr.Textbox(label="Status", lines=2)
    user question = gr.Textbox(label="; Ask a question about the PDF")
    ask_button = gr.Button(" Get Answer")
    answer_output = gr.Textbox(label=" Answer", lines=4)
    upload_button.click(fn=process_pdf, inputs=pdf_input,
outputs=[status_output, qa_chain_state])
    ask_button.click(fn=answer_question, inputs=[user_question,
qa chain state], outputs=answer output)
demo.launch()
```

first run ollama then run chatbot.py file:-



#### http://127.0.0.1:7860/



Here's a line-by-line explanation of your PDF Q&A chatbot using Gradio, LangChain, HuggingFace Embeddings, FAISS, and Ollama (LLaMA 3):



import gradio as gr

☐ Imports **Gradio** — the Python library to build interactive web UIs for AI apps.

from langchain community.document loaders import PyPDFLoader

→ Loads the PyPDFLoader to extract text from PDFs using LangChain's community-maintained tools.
from langchain.text_splitter import RecursiveCharacterTextSplitter
☐ Used to <b>split long PDF text into smaller chunks</b> (based on characters) so the model can handle it.
from langchain_huggingface import HuggingFaceEmbeddings
☐ Imports sentence embeddings from HuggingFace (used for semantic search).
from langchain_community.vectorstores import FAISS
→ FAISS is a fast vector database. It stores embeddings and enables efficient similarity search.
from langchain_ollama import OllamaLLM
→ Uses the <b>Ollama</b> local LLM interface to run llama3 or any other local model.
from langchain.chains import RetrievalQA
→ Combines a retriever (FAISS) + LLM (Ollama) to build a <b>Question Answering chain</b> .

## **PDF Processing Functions**

## 1. Load the PDF and split it

```
def load_pdf(path):
    loader = PyPDFLoader(path)
    docs = loader.load_and_split()
    return docs
```

- PyPDFLoader (path) reads the PDF.
- .load and split() reads and divides it into sections.
- Returns the list of document chunks (LangChain Document objects).

## 2. Turn documents into vector embeddings

```
def create_vectorstore(docs):
    splitter = RecursiveCharacterTextSplitter(chunk_size=500,
    chunk_overlap=50)
    texts = splitter.split documents(docs)
```

• Splits documents into 500-character chunks with 50-character overlap.

• Useful for preserving context across splits.

```
embeddings = HuggingFaceEmbeddings (model_name="sentence-transformers/all-MinilM-L6-v2")
```

• Loads a **small**, **fast** transformer model for text embeddings.

```
vectordb = FAISS.from_documents(texts, embedding=embeddings)
return vectordb
```

• Converts document chunks to vectors using embeddings and stores them in FAISS.

#### 3. Load the Ollama-based LLM

```
def load_llm():
    return OllamaLLM(model="llama3")
```

- Loads the LLaMA 3 model from Ollama.
- Works offline if Ollama and the model are installed locally.

### 4. Build RetrievalQA Chain

```
def create_qa_chain(llm, vectordb):
    retriever = vectordb.as retriever(search kwargs={"k": 3})
```

• Retrieves top 3 most relevant text chunks based on query.

```
return RetrievalQA.from_chain_type(llm=llm, chain_type="stuff",
retriever=retriever)
```

- Combines retriever + LLM into a pipeline:
  - (1) Retrieve relevant chunks  $\rightarrow$  (2) Ask LLM to generate answer from them)

## Gradio Callback Functions

### 5. Handle PDF Upload and Processing

```
def process_pdf(file_path):
    try:
    docs = load_pdf(file_path)
    vectordb = create_vectorstore(docs)
    llm = load_llm()
    qa_chain = create_qa_chain(llm, vectordb)
```

• Loads the PDF → splits it → creates vector index → loads the model → builds QA chain.

```
return f" ✓ PDF processed. Ready to ask questions.", qa_chain except Exception as e:
return f" X Error while processing PDF: {e}", None
```

• Returns status message + chain (stored in Gradio's gr.State()).

### 6. Handle Question-Answering

```
def answer_question(question, qa_chain):
    if not qa_chain:
        return "\( \Lambda \) Please upload and process a PDF first."
```

• Checks if QA chain is initialized.

```
if not question.strip():
    return "⚠ Please enter a question."
```

• Checks if question is empty.

```
result = qa_chain.invoke({"query": question})
return result["result"] if isinstance(result, dict) else result
```

• Uses .invoke() to get answer from chain and returns it.

## **Gradio UI**

#### 7. Build the interface

```
with gr.Blocks() as demo:
```

• Starts building a Gradio app.

```
gr.Markdown("#  Ask Questions About Your PDF")
gr.Markdown("Built with **Ollama + LLaMA 3 + HuggingFace + FAISS**")
```

• Shows title and description.

```
qa_chain_state = gr.State()
```

• Keeps the ga chain saved between button clicks.

```
pdf_input = gr.File(label=") Upload a PDF", file_types=[".pdf"])
upload_button = gr.Button(" Process PDF")
status output = gr.Textbox(label="Status", lines=2)
```

- File uploader for PDF.
- A button to trigger processing.
- A textbox to show processing result.

```
user_question = gr.Textbox(label=" Ask a question about the PDF")
ask_button = gr.Button(" Get Answer")
answer output = gr.Textbox(label=" Answer", lines=4)
```

- User enters question.
- Button to submit.
- Answer shown in textbox.

### 8. Link buttons to functions

```
upload_button.click(fn=process_pdf, inputs=pdf_input,
outputs=[status_output, qa_chain_state])
```

- When "Process PDF" is clicked:
  - o process\_pdf() is called with the uploaded file path.
  - o Returns status + QA chain (saved in state).

```
ask_button.click(fn=answer_question, inputs=[user_question,
qa chain state], outputs=answer output)
```

- When "Get Answer" is clicked:
  - o Sends user question and saved QA chain.
  - o Displays answer.

### 9. Launch the app

```
demo.launch()
```

• Starts the Gradio web server (by default: http://localhost:7860).

# **✓** Final Summary

This app does:

- 1. Load a PDF and split it into chunks.
- 2. Convert chunks into embeddings using HuggingFace.
- 3. Store in FAISS for similarity search.
- Use LLaMA 3 (via Ollama) to generate answers from retrieved chunks.
   Provide a UI to upload a PDF and ask questions about it.