Al Copilot for Renewable Energy Data Rooms - Project Plan

This plan outlines the steps to build a prototype based on the provided challenge description.

1. Core Goal:

- Build an AI assistant that allows users to upload renewable energy project documents (PDFs, etc.) and ask natural language questions.
- The assistant must provide answers backed by specific references (quotes, page numbers) from the uploaded documents.

2. Key Features (MVP - Minimum Viable Product):

- File Upload: Allow users to upload multiple PDF documents.
- Document Processing:
 - o Extract text content from PDFs.
 - Chunk the text into manageable segments.
 - Generate embeddings for each chunk.
- **Indexing:** Store chunks and their embeddings in a vector database (e.g., FAISS for simplicity initially).

• Querying (RAG):

- Accept natural language questions from the user.
- Embed the user's question.
- Retrieve relevant document chunks from the vector database based on the question embedding.
- Construct a prompt for an LLM, including the user's question and the retrieved chunks.
- Call an LLM API to generate an answer based only on the provided context.
- **Answer Presentation:** Display the LLM's answer along with the source document name and page number for the chunks used.
- User Interface: Basic UI (using Streamlit) for file upload, question input, and displaying results.

3. Technology Stack Considerations (Based on Hints):

- Document Parsing: PyPDF2 or PyMuPDF (simpler) or PDFPlumber (more structured).
- **Embeddings:** Sentence Transformers (e.g., all-MiniLM-L6-v2 runs locally) or an API like OpenAI's.
- Vector Database: FAISS (runs locally, good for prototypes) or cloud options like Pinecone/Weaviate.
- LLM: An API-based model (e.g., Google's Gemini API, OpenAI's GPT API) is

- usually easiest for hackathons.
- RAG Framework (Optional but Recommended): LangChain or LlamaIndex can simplify the pipeline orchestration.
- **UI:** Streamlit or Gradio.

4. Development Steps:

- Step 0: Setup: Create a project environment (e.g., using venv or conda), install initial libraries.
- Step 1: Document Loading & Chunking: Implement functions to read PDFs, extract text, and split into chunks.
- Step 2: Embedding & Indexing: Implement functions to generate embeddings for chunks and store them in FAISS.
- Step 3: RAG Core Logic: Implement the retrieval (query FAISS) and generation (prepare prompt, call LLM API) steps. Ensure source tracking.
- Step 4: Basic UI: Build a simple Streamlit interface to tie everything together (upload, input, output).
- Step 5: Testing & Refinement: Test with sample documents and questions.
 Refine chunking, prompting, and retrieval strategies.

5. Advanced Features (Post-MVP):

- Support for other file types (Excel, Word).
- More sophisticated chunking and metadata extraction.
- Handling very large data rooms (scalability improvements).
- Checklist auto-population feature.
- Clickable links in references (if UI allows).

Next Steps:

- Decide on the specific libraries for parsing, embeddings, vector store, and LLM.
- Start implementing Step 1: Document Loading & Chunking.