CareerCraft: A Smarter, Simpler Workspace for Your Job Search

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Why I Built CareerCraft

Large language models (LLMs) have rapidly transformed the landscape of artificial intelligence and human-computer interaction. From writing assistance to research acceleration, they are becoming integral to both personal and professional workflows. While much of the discussion around LLMs focuses on enterprise-scale applications or public-facing chatbots, their potential for solving highly personal productivity problems is equally profound. CareerCraft is an attempt to harness this potential through a privacy-conscious, locally hosted, AI-powered tool designed to streamline the job search process.

The idea for CareerCraft came from a deeply personal pain point. Currently in my job search, I find myself scattered across multiple systems, spreadsheets for application tracking, documents for interview stories, and websites for resume feedback. This fragmented approach is not only inefficient but also mentally taxing. I wanted a centralized solution that could help me organize, reflect, and improve my job application materials without the friction of constantly switching tools.

While there are tools in the market that offer resume review, interview preparation, or job tracking capabilities, most are either paid services or cloud-based platforms. Using them often requires handing over personal resumes, job descriptions, or sensitive application notes to external servers. In contrast, CareerCraft is built entirely as an on-premise application. It runs locally using a language model hosted via Ollama and provides complete data ownership to the user. This ensures that all career-related content, from resume drafts to personal stories, remains private while still benefiting from the intelligence of a modern model like llama 3.

CareerCraft integrates three core modules: a Resume Rewriter, a STAR Story Generator, and an Application Tracker. These components work together to form a personal productivity assistant tailored to job seekers. The application is built using Streamlit and demonstrates how LLMs can be incorporated into real-world workflows to solve everyday problems meaningfully and ethically.

How It Works: Inside the Tech Behind CareerCraft

CareerCraft is built as a modular, multi-page web application using Streamlit, where each page maps directly to a specific user workflow: resume rewriting, STAR story generation, and application tracking. The Resume Tool and STAR Generator both rely on a retrieval-augmented generation (RAG) setup that uses FAISS¹ for similarity search and the llama 3: instruct model (8B, quantized) running locally via Ollama for inference. This specific model was selected because it is optimized for instruction-following tasks, performs efficiently on consumer hardware, and excels at generating personalized content such as resumes and behavioral stories. The resume_examples.json file acts as the structured knowledge base. In contrast, the Application Tracker uses simple CSV-based persistence for lightweight, no-database job

logging. This hybrid architecture enables CareerCraft to deliver AI-assisted insights while preserving user privacy through fully local processing.

Resume Rewriter

The Resume Rewriter module is the most technically involved feature in the application. At its core, it implements a simplified RAG pipeline to rewrite or enhance resume content based on a specific job description. The system starts with a curated knowledge base stored in a JSON file that contains strong resume bullet points across multiple roles and domains. Each entry in this dataset is converted into vector embeddings using a local embedding model and stored using FAISS for fast similarity search.

When a user provides their resume and a job description, the system parses the inputs and constructs a query. It then retrieves the most relevant resume examples from the knowledge base and combines them with the user's content to build a context-augmented prompt. This prompt is passed to a llama 3 model running locally via Ollama. The generated output suggests improvements to the resume tailored to the job description and based on high-quality examples. This entire workflow happens locally, ensuring data security while still delivering context-aware suggestions.

STAR Story Generator

The STAR Story Generator module is designed to help users craft effective behavioral interview responses using the Situation-Task-Action-Result (STAR) format. The interface allows users to input stories manually and label them by theme (such as teamwork, leadership, or conflict resolution). These stories are then displayed in the interface for quick reference and can optionally be enhanced using the model by incorporating resume or job description context. This functionality makes the tool especially helpful for practicing behavioral interview responses in a structured and personalized way.

Application Tracker

The Application Tracker serves as a lightweight database for managing job applications. It uses a simple CSV file to persist data such as job title, company name, application source, date applied, current status, and notes. Users can view, update, and delete entries through an interactive interface. Though simple, this tracker provides an essential structure for managing what is often an overwhelming part of the job search process.

All three modules are integrated within a single Streamlit app using its multi-page support. The interface is clean, intuitive, and designed for use on a local machine. No internet connection is required after the initial model setup, reinforcing the privacy-first design philosophy of CareerCraft.

What I Learned and What I Struggled With

Building CareerCraft offered a hands-on opportunity to explore multiple aspects of applied LLM development. It deepened my understanding of how local llama 3 models differ from cloud-hosted APIs in terms of performance, latency, and prompt behavior. Working with Ollama and the llama 3 model (8B parameters, quantized, instruction-tuned) allowed me to experiment with local inference capabilities while

staying entirely within the privacy constraints I had set for myself. The experience also taught me how to manage embeddings and retrieval systems for effective RAG pipelines, which was particularly important in the Resume Rewriter module.

The project also helped me appreciate the user experience and interface design challenges involved in building a real application. Managing state in Streamlit, especially across multiple pages and dynamic elements, required careful structuring of inputs and caching. Building useful prompts that balance structure with flexibility was another key learning area, especially since resume editing and STAR storytelling are highly personalized tasks.

There were several technical and design challenges along the way. One of the main difficulties was optimizing the performance of local inference. Although running llama 3 locally is feasible, prompt length and embedding lookups must be handled carefully to avoid slowdowns. Another issue was data persistence. Because the app is meant to be entirely local, I opted for JSON and CSV files instead of cloud-based databases. This required building lightweight CRUD logic within the Streamlit app itself. Prompt engineering was also a challenge. I experimented with multiple strategies to keep the model outputs relevant and specific without making the responses sound robotic or too generic.

What's Next for CareerCraft

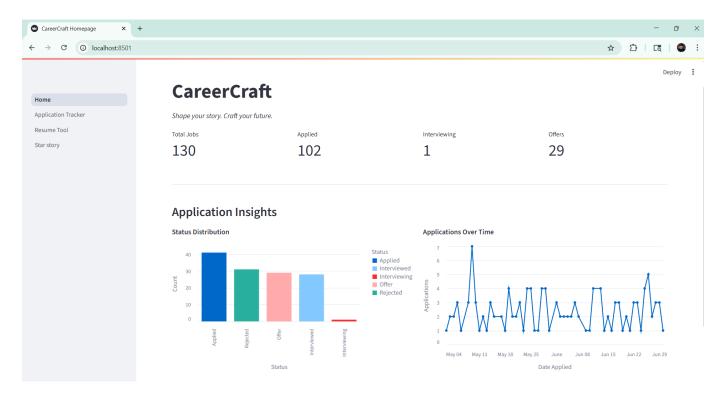
Looking ahead, there are several directions in which CareerCraft could evolve. One obvious extension is cloud deployment with optional authentication. This would allow users to access their job data across devices while still maintaining a strong privacy boundary. Another useful addition would be a more dynamic dashboard view. While CareerCraft already includes an overview page that displays a summary of active applications, enhancing it to show visual summaries of statuses, interview progress, and response rates would provide even greater clarity and insight into the user's job search journey. I also plan to expand the RAG knowledge base with role-specific resume examples and refine the embedding pipeline to use more domain-tuned models. Finally, scraping or importing jobs from platforms like LinkedIn could automate part of the tracking workflow and make the tool even more useful.

CareerCraft is more than just a demo of LLM capabilities. It is a reflection of how AI can be applied meaningfully to personal productivity challenges when built with intention, transparency, and user control. By bringing together local inference, retrieval-based generation, and a practical frontend, CareerCraft empowers job seekers to shape their story and craft their future.

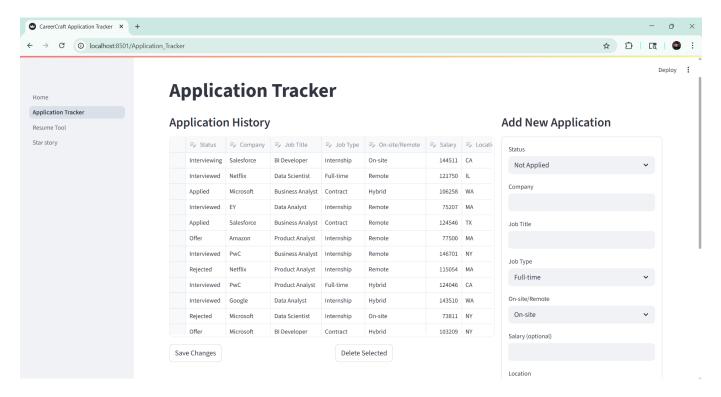
Appendix

Peek into CareerCraft's different pages:

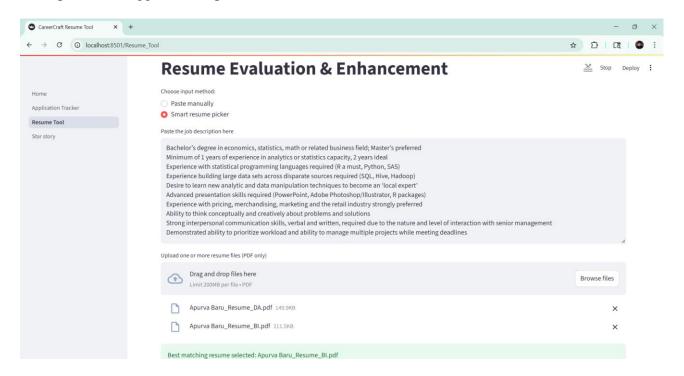
1. Homepage: The landing page that show the overview related to the applications that the user has did till date

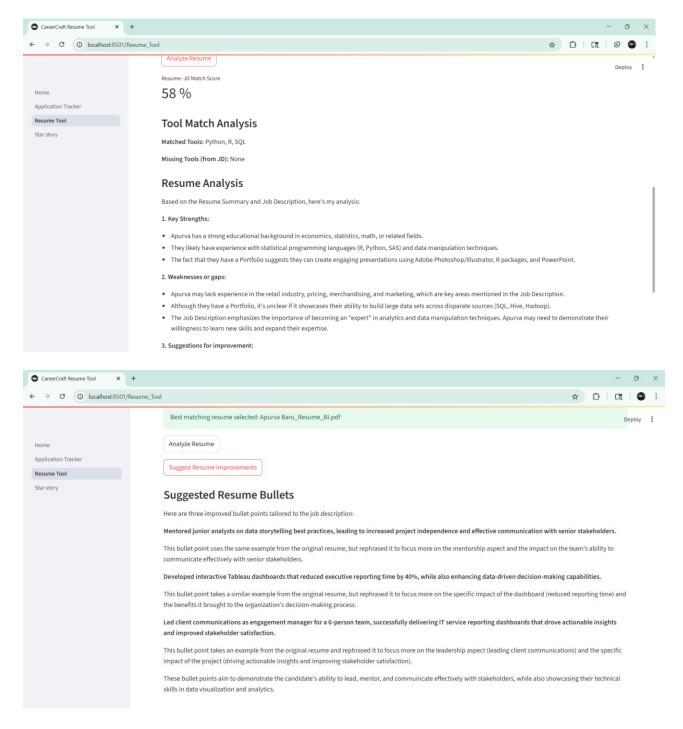


2. Application Tracker: User can add new jobs that they have applied to, update the status of previous jobs and delete the entries if required.



3. Resume Tool: Allows the user to match their resume to a job description, suggest changes, highlight strengths and weaknesses, help to choose which is the best resume as per the job description and suggest bullet points that the user can add.





4. Star Story generator: Helps user to generate a structured story based on their experience, role and resume

