



GLOBAL RESEARCH IMMERSION PROGRAM FOR YOUNG SCIENTISTS



CerebroCipher: Your Private AI Assistant in the Box

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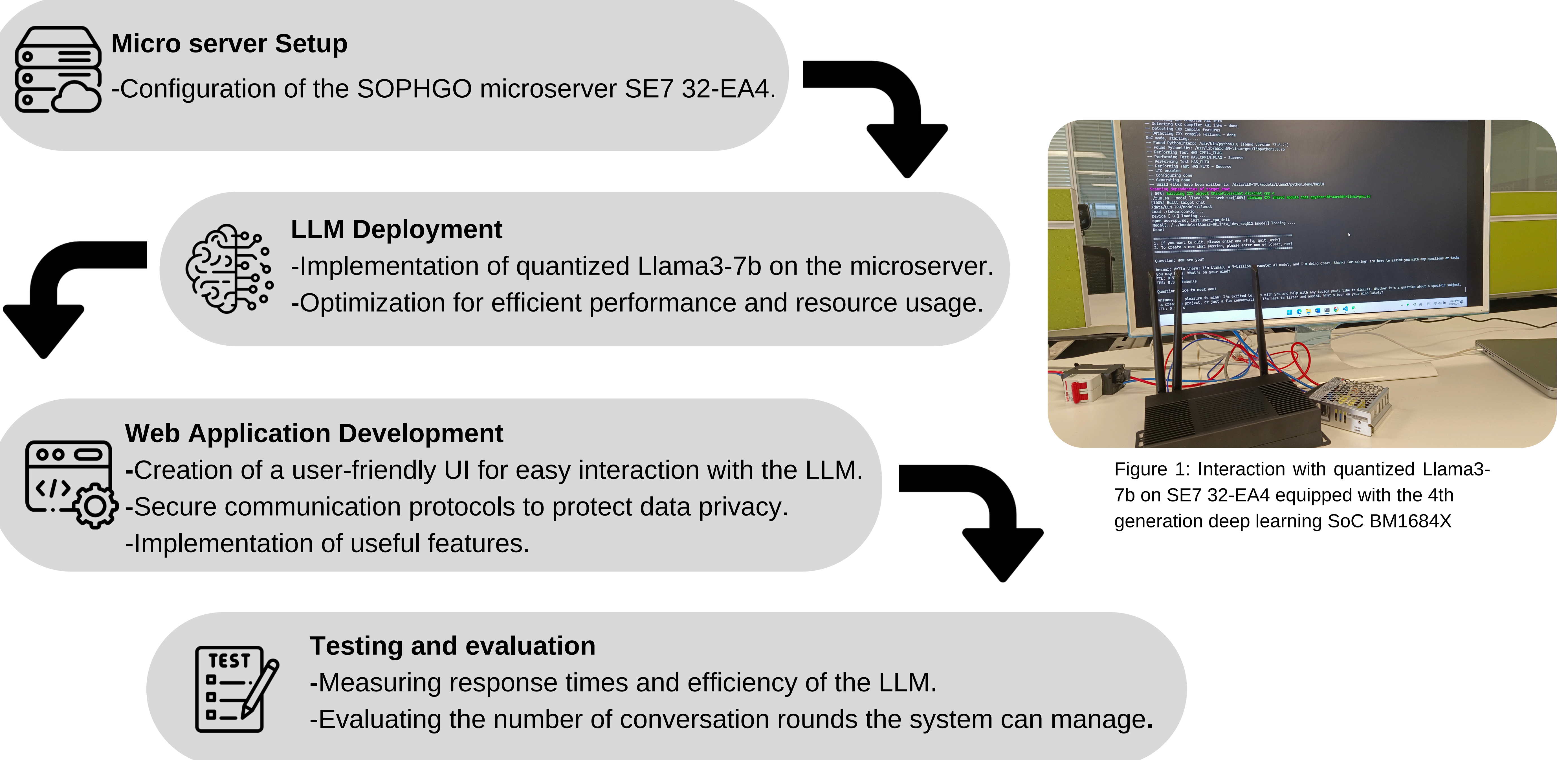
Background

Hosting Large Language Models(LLM) on traditional cloud servers causes significant memory and computational burdens, along with challenges of latency and privacy breach. This research addresses these challenges by using **edge computing solutions to deploy Llama3-7b on a micro server, providing LLM services for small and medium-sized enterprises (SMEs).**

Research Objectives

- Address privacy concerns by processing data locally.
- Deploy quantized Llama3-7b on edge servers which reduces memory and computational burdens.
- Utilize edge computing to improve performance, reduce latency, and lower data transfer costs for LLM services
- Provide small and medium-sized companies with a private and secure AI solutions.

Methods



Results

A web application was built with Gradio (Figure 2). Users can access it by simply entering the server address into their browser. The application supports text processing capabilities with short response times. An authentication page ensures that only permitted users can access the application (Figure 3). On average, it takes **0.773 seconds** for Llama3-7b to process and start generating responses, with an average processing rate of **8.078 tokens per second** (Figure 4). This has proven to be a stable and efficient AI solution with low latency.

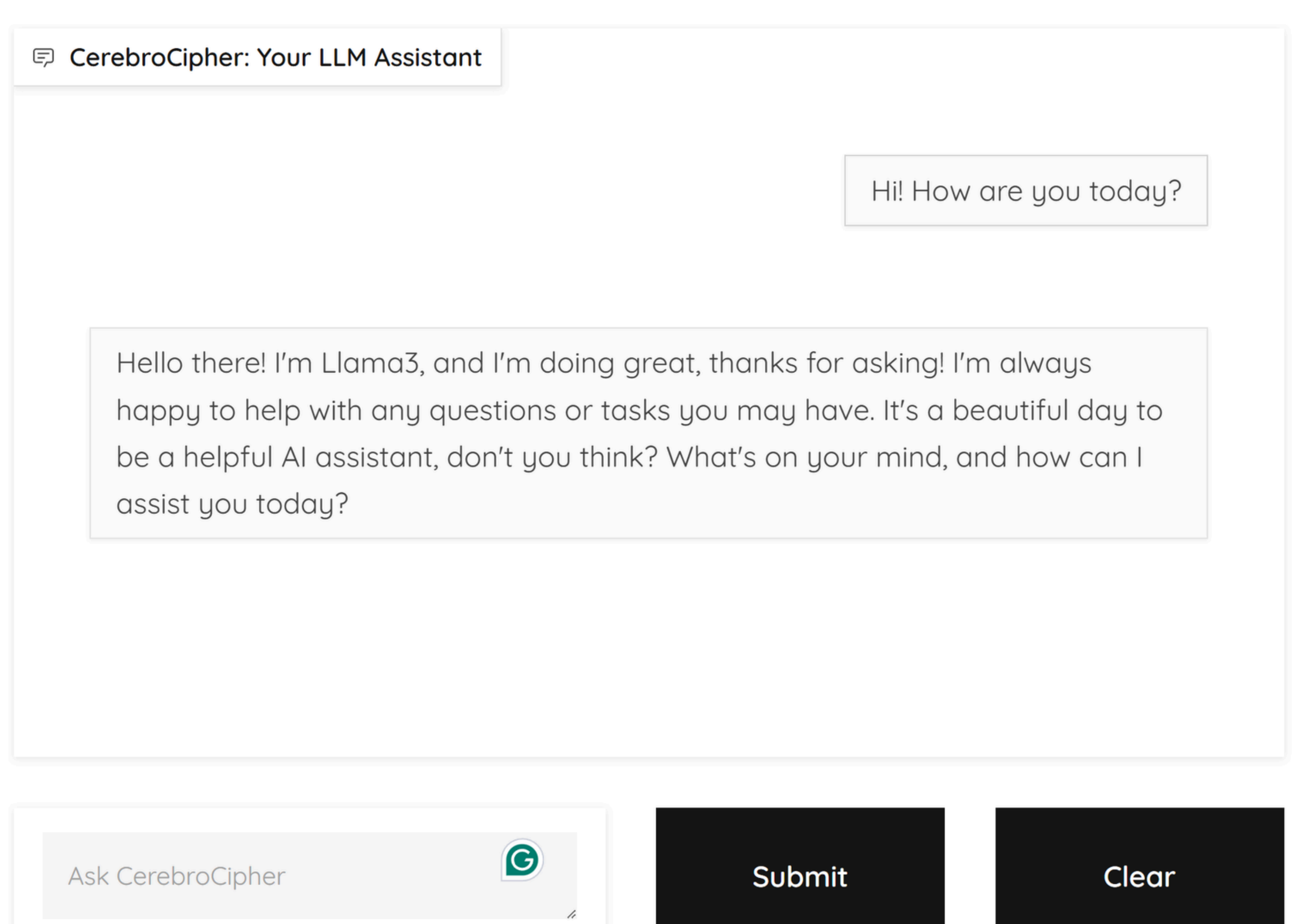


Figure 2: CerebroCipher web application

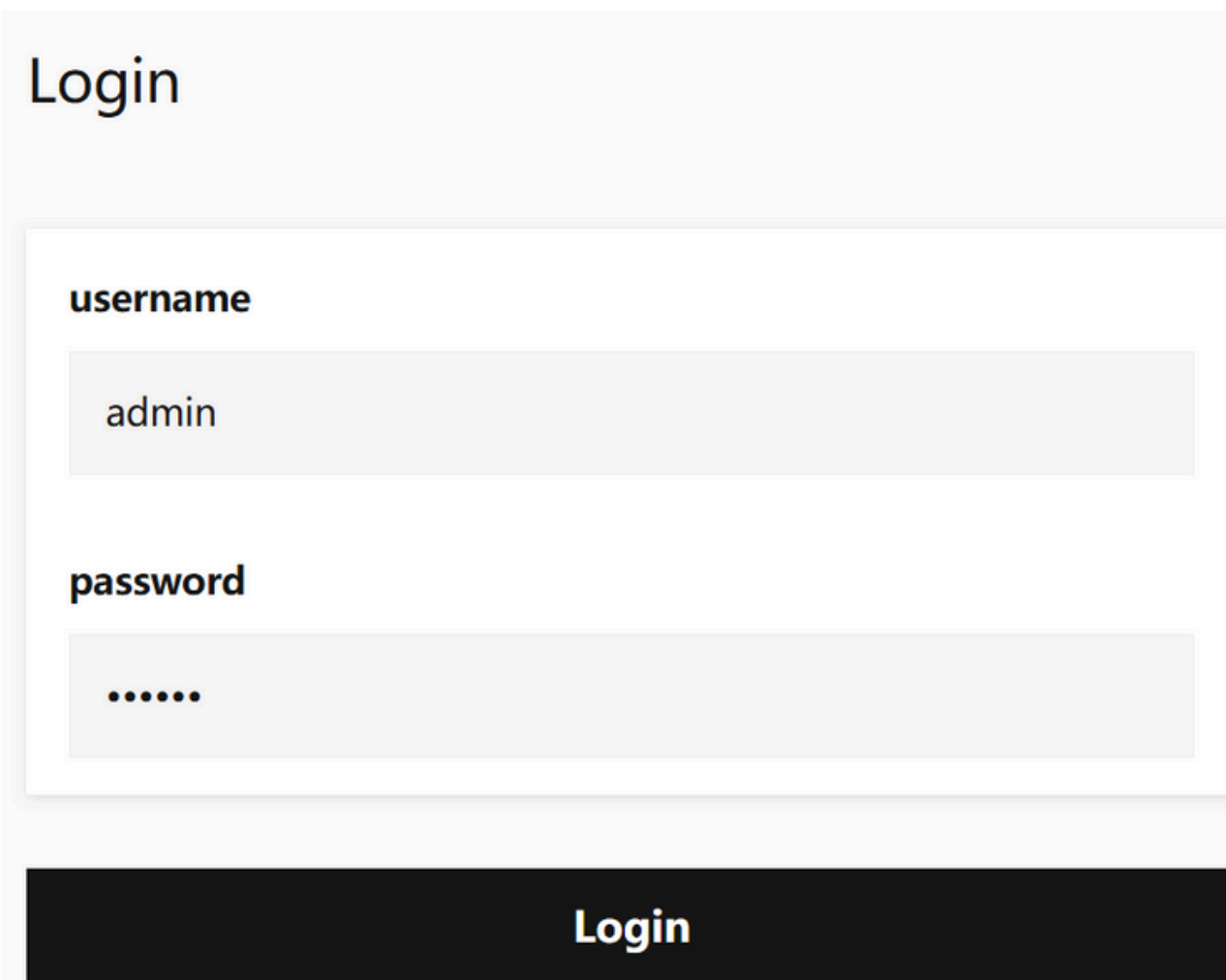
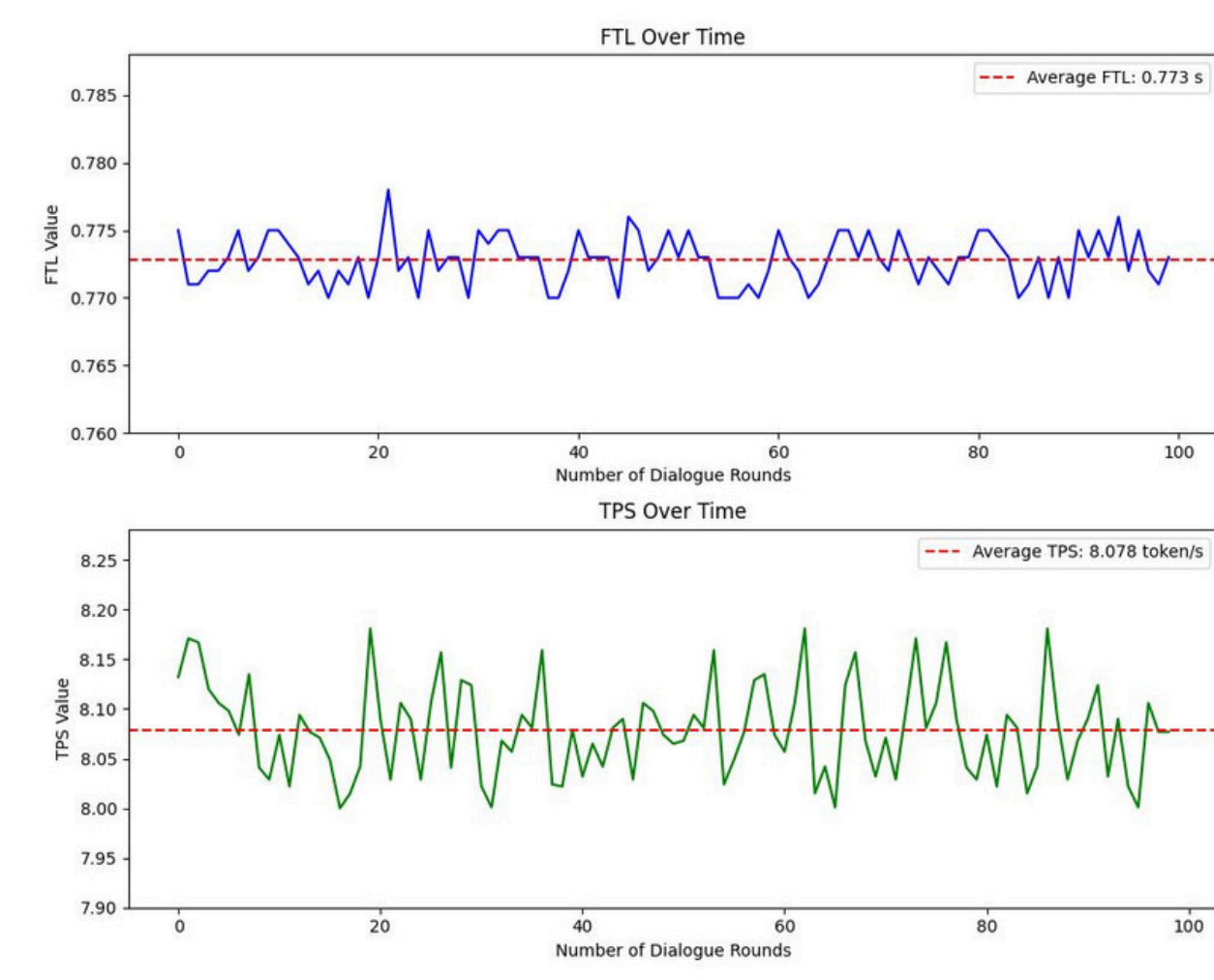


Figure 3: Authentication page



Discussion

In the current implementation, a constraint of limited memory from only one server was encountered, which impacts performance and scalability of our project. It restricts the maximum SEQLEN, feature complexity and necessitates single-thread usage, causing users to queue for responses. To address this issue, future improvements could include **deploying a stack of servers to handle more simultaneous users or exploring a device-server hybrid inference strategy to optimize performance and enhance user experience.**

Conclusions

A secure and efficient AI solution for small and medium-sized enterprises was successfully built by deploying Llama3-7b on SOPHGO SE7 microservers. The system supports fast text processing, providing robust capabilities while ensuring data privacy and control. Despite the current constraints on server memory and performance, the implementation demonstrates feasibility and potential of using edge servers to address privacy concerns associated with traditional cloud-based AI solutions. Future improvements will focus on enhancing scalability and performance to further meet the needs of SMEs.

Key References

Bang, J., Lee, J., Shim, K., Yang, S., & Chang, S. (2024, June 11). Crayon: Customized On-Device LLM via Instant Adapter Blending and Edge-Server Hybrid Inference. ArXiv.org. <https://doi.org/10.48550/arXiv.2406.07007>

Gradio. (n.d.). Gradio.app. <https://www.gradio.app/>

sophgo/LLM-TPU. (2024, July 31). GitHub. <https://github.com/sophgo/LLM-TPU>

Yao, Y., Duan, J., Xu, K., Cai, Y., Sun, Z., & Zhang, Y. (2024). A survey on Large Language Model (LLM) security and privacy: The Good, The Bad, and The Ugly. High-Confidence Computing, 4(2), 100211. <https://doi.org/10.1016/j.hcc.2024.100211>

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