



Zepto Chatbot – AI-Powered Grocery Assistant

By Fiza G | Quality Engineer



Project Overview

I created the Zepto Chatbot as a **self-learning project** to strengthen my technical profile beyond manual testing and explore **AI-driven automation concepts**.

My goal was to design a smart chatbot that mimics Zepto's customer assistant — answering grocery-related queries, checking FAQs, suggesting offers, and providing fallback AI-based responses when data isn't available.

This project started purely as a learning experiment, but it turned into a working, fully deployed chatbot with a clean UI, real-time interaction, and analytics.



Live App Link: <https://zepto-chatbot-demogit-ajbvzf5zvv975bhfzxth6s.streamlit.app/>



Objective

Coming from a QA background, I wanted to showcase that I could:

- Design and test complete applications — from logic to UI.
 - Integrate **AI models**, manage **APIs**, and handle **automation workflows**.
 - Build something functional that highlights curiosity and real implementation skills.
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Tech Stack

- **Language:** Python
 - **Frontend:** Streamlit
 - **Backend Logic:** Python functions (rule-based + AI fallback)
 - **AI Model:** LLaMA 3.2 via Hugging Face API
 - **Data Management:** JSON, CSV, Pandas
 - **Libraries Used:** rapidfuzz, huggingface_hub, dotenv, pandas, streamlit
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Architecture & Flow

1. **User Interface:** Built with Streamlit to create a simple, mobile-friendly chat experience.
 2. **Input Processing:** Cleans and normalizes user messages using Regex.
 3. **FAQ & Product Matching:**
 - Uses *RapidFuzz* for fuzzy matching to handle typos or similar phrases.
 - Checks local JSON data (chat_data.json, zepto_data.json) for product info and prices.
 4. **Festival Module:** Displays active offers or greetings if the date matches any event.
 5. **AI Fallback:**
 - Initially used **Hugging Face Inference API**, but later improved with **LLaMA** for more accurate and faster responses.
 - Helps answer new or unlisted queries naturally.
 6. **Logging:**
 - All user-bot conversations are saved in **JSON** and **CSV** formats.
 - This data is used later for dashboard analytics.
 7. **Dashboard (Streamlit Page):**
 - Tracks daily chat volume, frequent words, and last 20 messages.
 - Gives insights into user behavior and chatbot usage.
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Tools and Why I Used Them

Tool	Purpose	Why I Chose It
Python	Main programming language	Easy to integrate APIs and handle logic cleanly
Streamlit	Web interface & dashboard	Simple, interactive, and no HTML/CSS required
RapidFuzz	Text similarity	Handles spelling mistakes and fuzzy matches faster than fuzzywuzzy
Hugging Face / LLaMA	AI response generation	Free and lightweight alternative to OpenAI for testing
dotenv	Secret key management	Keeps API keys safe instead of hardcoding them
Pandas	Chat log storage and analytics	Makes CSV management and visual charts easy
JSON	FAQ & data storage	Lightweight and readable, no database setup needed

Testing Approach

As a QA Engineer, I applied testing concepts even while developing:

- **Functional Testing:** Validated all user message flows and responses.
- **Negative Testing:** Checked how the bot handles incomplete or irrelevant questions.
- **Regression Testing:** Ensured previous logic (FAQ, item checks) still worked after adding AI.
- **UI Testing:** Verified layout, message alignment, and responsive design in Streamlit.
- **Data Validation:** Cross-checked JSON data loading, file creation, and logging accuracy.

This helped me simulate both **QA and developer roles** in one project — from design to deployment.

Deployment

- Project hosted on **GitHub**.
 - Deployed on **Streamlit Cloud** using GitHub integration.
 - Streamlit automatically pulls updates from the repository on every commit.
 - All data and logs remain local to ensure privacy.
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Key Features

- Real-time chat interface with human-like responses.
 - Handles FAQs, products, and festival offers.
 - AI fallback using LLaMA model.
 - Automatic chat logging in JSON & CSV.
 - Password-protected admin view for logs.
 - Streamlit dashboard for analytics & word frequency charts.
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Challenges & Fixes

- **Slow and irrelevant responses** from the first Hugging Face API version → switched to **LLaMA model** for better accuracy and speed.
 - **Session timeout** issues in Streamlit → optimized with state handling and smaller wait time.
 - **Data mismatches** → used fuzzy logic (RapidFuzz) to improve matching accuracy.
 - **UI freezing** → added controlled rerun mechanism in Streamlit for smooth refresh.
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Dashboard Insights

The dashboard page gives:

- Total message count
- Unique user count
- Daily usage trends
- Most common query words

- Last 20 user messages (real-time table view)

This feature shows my ability to not only build, but also **analyze and test user data behavior** — a key QA mindset.

Drawbacks & Limitations

Even though the chatbot performs well for a demo project, it has a few practical limitations:

- **Streamlit Sleep Mode:** Since it's hosted on Streamlit Cloud's free tier, the app automatically goes to sleep after inactivity. This can delay the first response when it restarts.
- **Limited Session Memory:** The chatbot doesn't store long-term context; once the page refreshes, chat history resets.
- **Response Time Variability:** LLaMA API sometimes responds slower or gives off-topic answers depending on server load.
- **Static Data:** Product and festival data are stored locally in JSON files, so updates must be done manually instead of fetching live data.
- **No Real Database:** All logs are saved locally (CSV/JSON), which limits scalability if many users interact at once.
- **Basic UI Flow:** The design focuses on functionality; advanced features like user login, voice input, or real-time updates are yet to be added.
- **Limited Error Handling:** Network or API failures aren't fully handled with retry logic or fallbacks yet.



Learning Outcomes

- Strengthened my skills in **Python automation, AI integration, and UI testing**.
- Learned to handle **API errors, data parsing, and state management**.
- Improved my understanding of **end-to-end testing** in AI-driven applications.
- Built confidence in explaining complete workflow logic during technical interviews.



Future Improvements

- Add **database (SQLite or Firebase)** instead of local files.
- Train chatbot with custom Zepto-like data for higher accuracy.

- Introduce **voice input** or **multilingual support**.
 - Connect with **real Zepto APIs** for live inventory and offers.
 - Add **user authentication** for personalized interaction.
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About Me

I'm **Fiza G**, a **Quality Engineer** with 3.7 years of experience in **manual and automation testing** (Selenium, Python, Robot Framework, API Testing).

I'm passionate about building reliable, testable, and user-friendly systems.

This project is a reflection of my curiosity, self-learning approach, and interest in combining QA with AI and automation.

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