

E-Commerce Chatbot using GenAI

Fractal-Nexus Hackathon (March 2025)

Prince Kumar Singh (prince.singh@fractal.ai)

Overview

Customers often face difficulties due to complicated return policies, long processing times, and poor customer service experiences. Leverage GenAI to build an intelligent chatbot that:

- Guides users through the return/exchange process based on product type, policy, and purchase history.
- Understands customer intent (return, exchange, refund, replacement) and provides instant solutions.
- Checks order eligibility for returns/exchanges based on provided policy documents.
- Offers personalized recommendations for alternative products.
- Estimates customer mood and escalates issues accordingly.

Data Understanding

Two main source of data coming from other systems:

- Policy and rules related documents.
- Customer orders data.

Data Sources

Policy Info	Customer Orders Data
<ul style="list-style-type: none">● Return Window: 30-day return period from the delivery date.● Eligibility Conditions: Items must be in their original condition with tags and packaging.● Non-Returnable Items: Final sale, custom-made, and used items.● Process Steps: Initiation via chatbot, Return Authorization (RA) number, and trackable shipping.	<ul style="list-style-type: none">● Order ID: Unique identifier for each purchase.● Product Category: Specifies the type of product purchased.● Purchase Date: Helps determine return eligibility based on policy timelines.● Return/Exchange Status: Indicates whether an order has been returned, exchanged, or refunded.● Customer ID: Links orders to specific users for personalized interactions.

Parsing Policy Data

- The documents are broken down into sections and subsections.
- Each of the subsections are summarized into key points and their main intents are derived to create a better structured format.

```
[
  {
    "intents": [
      "return",
      "exchange"
    ],
    "summary": [
      "There is a 30-day return or exchange policy.",
      "Items must be unworn with tags attached.",
      "Original packaging is required for returns.",
      "A valid proof of purchase is needed."
    ]
  },
  {
    "intents": [
      "return",
      "refund"
    ]
  }
]
```

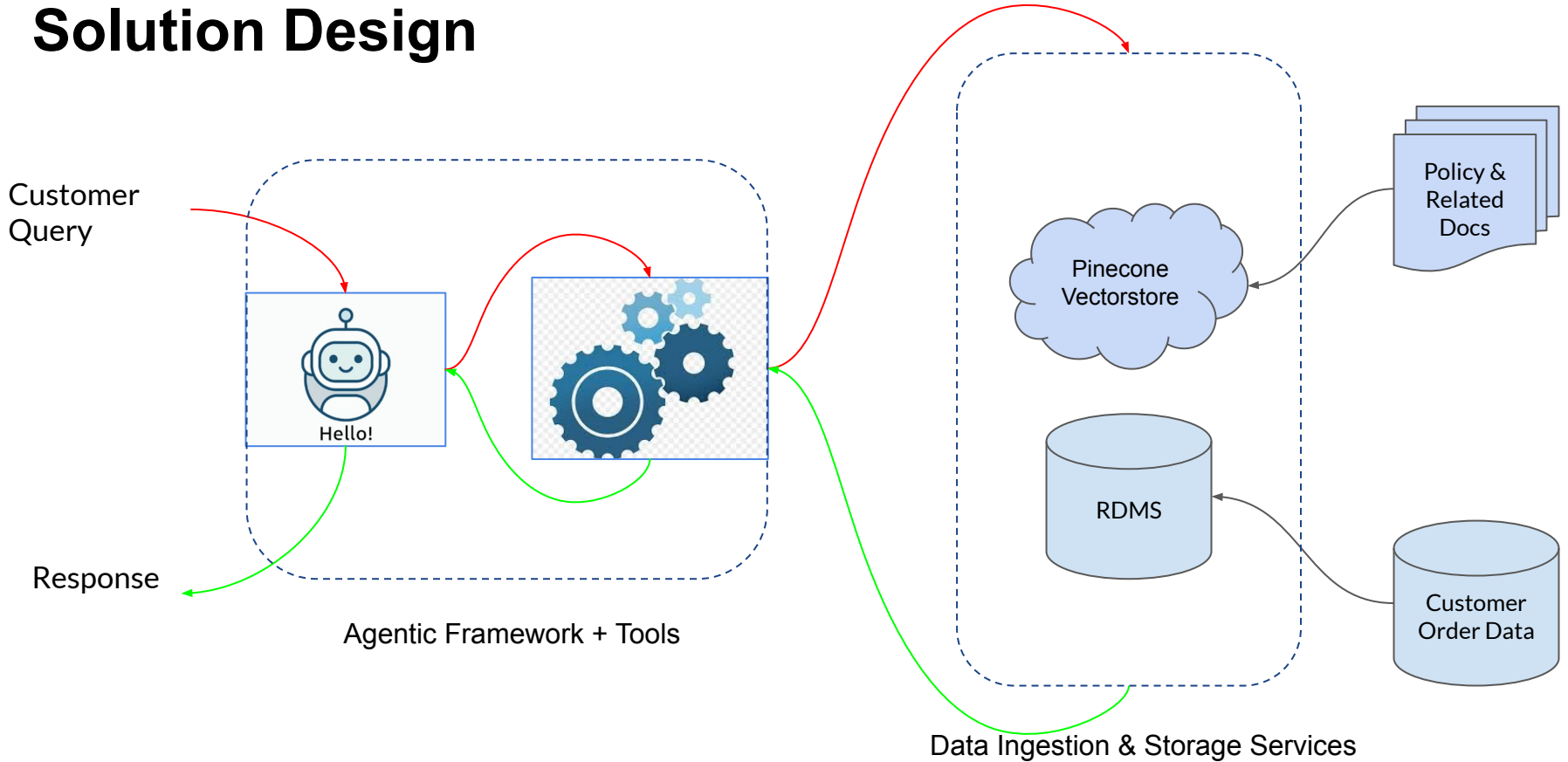
Policy Data Ingestion Service

- The following fields are created for each sentence in the JSON file:
 - id: This a unique ID for each of the unique sentence
 - metadata: This field stores the text and list of intents applicable to the text
 - values: Embedding vector of each sentence. Here **all-mpnet-base-v2** is used.
- All the records are saved in Pinecone vector store
- This service is always live and checks for any new document to process every hour.
- Once the document is processed it saves the filename in database.

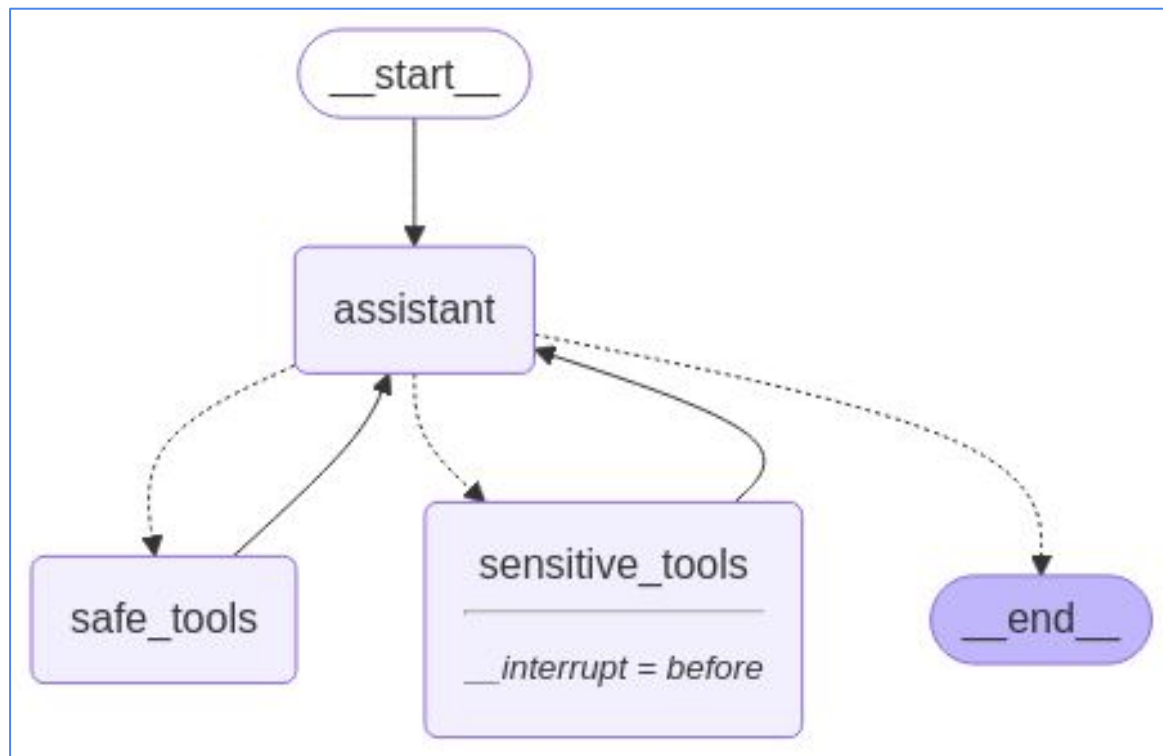
Orders Data Ingestion Service

- This service checks for new records in the given physical file location.
- It uses Order ID as unique key to identify new orders data.
- It standardizes the columns names and date format and saves data to database.
- This service checks for new records every hour.

Solution Design



Bot Graph



Tools Used

- **Product-Recommendor-By-OrderID**
 - Gets products similar to the product given any order ID.
 - Checks for same product category, size and gender (or Unisex)
- **Generate-Return-Authorization**
 - Generates a unique return authorization number for a given order ID.
 - Considered as “Sensitive” tool in the model.
- **Get-Order-Details**
 - Fetches order details from the database given order ID.
- **Get-Relevant-Policies-By-Query**
 - Retrieves relevant policy details given an order ID.
 - Fetches policies from vector store by similarity and intents metadata
 - Re-ranks the policies using bge-reranker-v2-m3 reranker model and filters all having score lower than threshold
- **Days-Since-Date**
 - Calculates the number of days passed since the given order date.

Tech Stack Used

- UI/UX
- Chatbot and Tools for RAG framework
 - Langgraph - very granular control of data flow, flexible, less black-box and more developer friendly
- LLM
 - gemma2-9b-it from Google - lightweight, state-of-the-art open model
 - llama3-70b-8192 from Meta - state-of-the-art open LLM model
 - Groq API - free API
- Vectorstore
 - Pinecone - free, serverless, low latency, reranking and metadata filtering available
 - Embedding - sentence-transformers/all-mpnet-base-v2 through Huggingface - well-established, pre-trained, suitable for similarity tasks for RAG like applications

Next Steps

- Fine-tuning and further prompt-engineering for better accuracy and response.
- Add a gateway node which can act as filter for all kind of rules.
- Add multi-user support and session handling.
- Design multi-agent system rather than can enable all the models to run with smaller scope of work and lightweight models thus decrease running cost.
- Make the agent interact with other systems like - recording return requests, informing inventory and shipping partners in automated way.

Thank You