

Assignment 6. Conversational Q&A using AzureOpenAI

Title: - Conversational Q&A System using Azure OpenAI and DockerHub

Objective

The objective of this assignment is to design and implement a Conversational Q&A system. Participants will gain hands-on experience with Azure OpenAI for embeddings, LangChain for chunking and orchestration, and FAISS for vector-based retrieval. They will also learn to build an interactive Streamlit UI, deploy it on Azure App Service, and finally dockerize the application for publishing on DockerHub.

This exercise simulates a real-world scenario where organizations use domain-specific datasets and large language models to build intelligent, searchable, and conversational applications.

Conversational Q & A Fitness Document

The chosen dataset is a comprehensive collection of specifications of Indian fitness plans and exercises across various brands. It includes attributes such as:

- Current Weight (kg)
- BMI
- Health goal (e.g., weight loss, muscle gain, maintenance)
- Health condition (e.g., diabetes, heart issues)
- Activity preference (e.g., yoga, cardio, strength training)
- Other relevant health details

This dataset is ideal for a conversational Q&A system because it contains structured attributes that naturally align with user queries.

Examples of queries:

- “Show me fitness plans and exercises diet plan for weight loss with diabetes-friendly meals.”

This dataset represents a structured fitness and wellness knowledge base, which makes it useful for building a virtual fitness plan or exercise consultant.

Real-world applications include:

- Online fitness plan or exercise recommendation chatbots for gyms or fitness centers
 - AI-powered comparison tools for fitness and wellness websites
 - Customer service assistants for nutritionists or fitness trainers
 - Decision-support tools for prospective users
-

Assignment Tasks

Task 1: Dataset Selection & Description

For this project, I selected an Indian fitness plans and exercises dataset containing attributes like Age, Gender, Weight (kg), Height (m), Max_BPM, Avg_BPM, Resting_BPM, Session_Duration (hours).

This dataset is well-suited for a Q&A system because it allows users to search for fitness plans and exercises based on multiple attributes simultaneously and supports intelligent recommendations.

Such a system can help prospective users quickly find the best fitness plan or exercise that matches their preferences, reducing the need for manual filtering on fitness plan or exercise websites.

```
# Load gym.csv
if os.path.exists("gym_members_exercise_tracking.csv"):
    df_gym = pd.read_csv("gym_members_exercise_tracking.csv")
    for _, row in df_gym.iterrows():
        text = " ".join([f"{col}: {row[col]}" for col in df_gym.columns if pd.notna(row[col])])
        docs.append(Document(page_content=text))

if os.path.exists("megaGymDataset.csv"):
    df_gymdataset = pd.read_csv("megaGymDataset.csv")
    for _, row in df_gymdataset.iterrows():
        text = " ".join([f"{col}: {row[col]}" for col in df_gymdataset.columns if pd.notna(row[col])])
        docs.append(Document(page_content=text))
```

Task 2: Data Chunking & Embedding Generation

- Used **LangChain** to convert each row of the dataset into a text-based document.
- Generated embeddings using **Azure OpenAI's text-embedding-3-small** model.
- Stored embeddings locally to enable fast semantic search during queries.

```
endpoint = os.getenv("ENDPOINT_URL", "https://susmi-mfowdoli-eastus2.openai.azure.com/")
deployment = os.getenv("DEPLOYMENT_NAME", "gpt-4.1-nano")
subscription_key = os.getenv("AZURE_OPENAI_API_KEY", "Ezgggca8H0eO8t3klmVgAYDcTRZrZ9nicV")

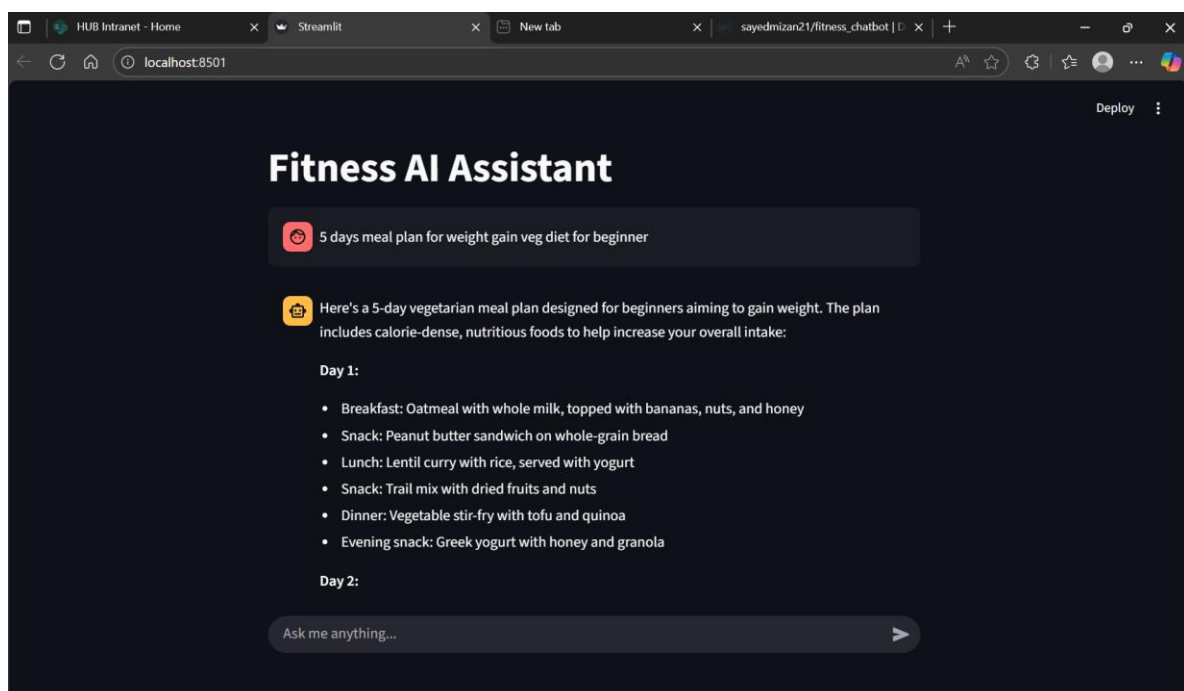
# Initialize Azure OpenAI client with key-based authentication
client = AzureOpenAI(
    azure_endpoint=endpoint,
    api_key=subscription_key,
    api_version="2025-01-01-preview",
)
```

Task 3: Build FAISS Index

```
azure_embeddings = AzureOpenAIEmbeddings(  
    azure_deployment="text-embedding-3-large",  
    model="text-embedding-3-large",  
    api_key="Ezgggca8H0e08t3klmVgAYDcTRZrZ9nicVWjQaz0jEOuaJtvpoyIJQQJ99BIACHYHv6XJ3w3AAAAACOGYVH",  
    azure_endpoint=endpoint,  
    api_version="2024-12-01-preview",  
)
```

Task 4: Create Streamlit UI

```
st.title("Fitness AI Assistant")  
  
if "messages" not in st.session_state:  
    st.session_state.messages = []  
  
for message in st.session_state.messages:  
    with st.chat_message(message["role"]):  
        st.markdown(message["content"])  
  
if prompt := st.chat_input("Ask me anything..."):  
    st.session_state.messages.append({"role": "user", "content": prompt})  
    with st.chat_message("user"):  
        st.markdown(prompt)  
  
    with st.chat_message("assistant"):  
        message_placeholder = st.empty()  
        full_response = ""  
        reply = Retrieval_chain(prompt)  
        full_response = reply['result']  
        message_placeholder.markdown(full_response)  
    st.session_state.messages.append({"role": "assistant", "content": full_response})
```



Task 5: Dockerization & Publishing on DockerHub

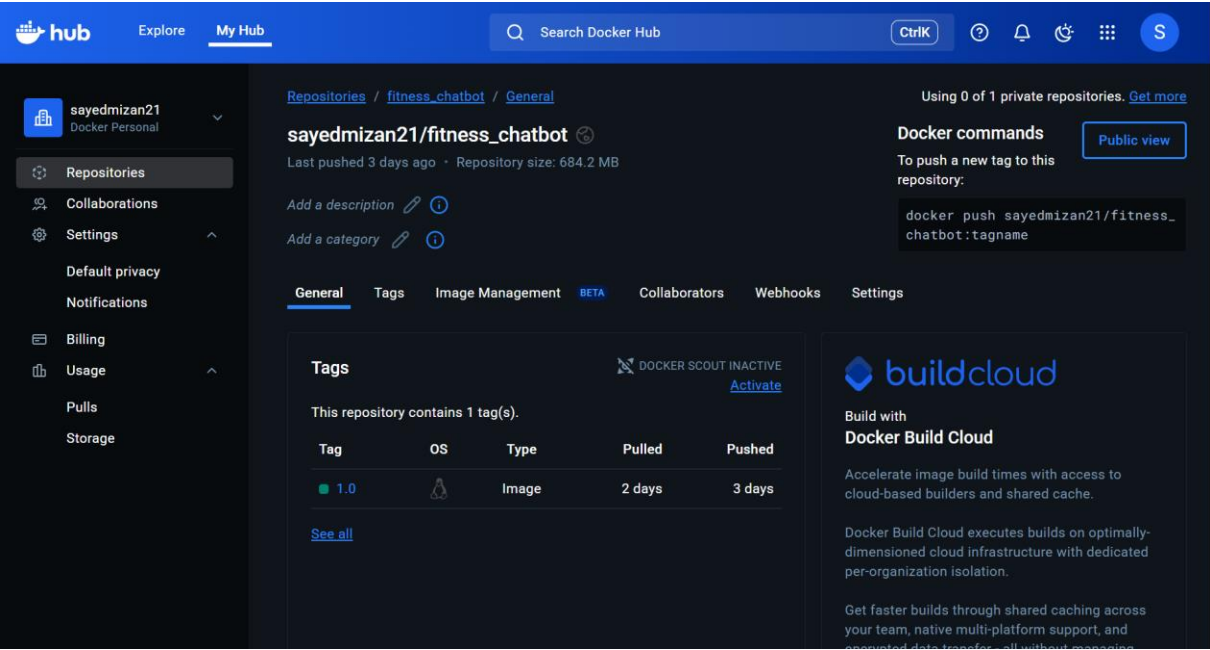
Dockerfile

```
Dockerfile
1  # Use official Python runtime as a parent image
2  FROM python:3.10-slim
3
4  # Set environment variables
5  ENV PYTHONDONTWRITEBYTECODE=1
6  ENV PYTHONUNBUFFERED=1
7
8  # Set working directory
9  WORKDIR /app
10
11 # Install system dependencies
12 RUN apt-get update && apt-get install -y \
13     build-essential \
14     && rm -rf /var/lib/apt/lists/*
15
16 # Copy requirements.txt and install Python dependencies
17 COPY requirements.txt /app/
18 RUN pip install --upgrade pip
19 RUN pip install -r requirements.txt
20
21 # Copy the rest of the application code
22 COPY . /app/
23
24 # Expose the port streamlit runs on
25 EXPOSE 8501
26
27 # Run the Streamlit app
28 CMD ["streamlit", "run", "app.py", "--server.port=8501", "--server.address=0.0.0.0"]
29
```

Requirements file

```
requirements.txt
1  streamlit
2  pandas
3  numpy
4  langchain
5  langchain-openai
6  faiss-cpu
7  openai
8  langchain-community
9  tiktoken
```

Docker Hub



Docker Desktop

