Case Analysis Assistant in Danswer Application

The goal of this project was to integrate a Case Analysis Assistant feature in the Danswer application. This assistant is designed to analyze legal case queries and provide a prediction (e.g., Accepted/Rejected) along with reasoning for the prediction. The implementation involved both backend and frontend work, as well as modifications to configuration files to integrate FastAPI, Hugging Face models, Docker, and the application's assistant-handling logic.

Server Implemented on (danswer):

(http://ec2-54-79-231-211.ap-southeast-2.compute.amazonaws.com)

Server Referenced (open-webui-main):

(http://ec2-13-237-148-39.ap-southeast-2.compute.amazonaws.com)

(Site: http://54.79.231.211)

1) open-webui-main server:

Path: cd open-webui-main/backend/apps/caseprediction/:

Files present in caseprediction dir:

→ <u>abc.py, config.json, main.py, model.safetensors, special_tokens_map.json, spiece.model, tokenizer_config.json</u>

Summary: The case prediction logic from the open-webui-main server served as the basis for this implementation. The caseprediction directory on the open-webui-main server provided a framework for handling case queries using AutoModelForSequenceClassification and AutoTokenizer, as well as generating reasoning with HuggingFaceHub.

Key components:

- The abc.py file in the open-webui-main server used SentenceTransformer for encoding queries and HuggingFaceHub for reasoning generation (NousResearch/Nous-Hermes-2-Mixtral-8x7B-DPO).
- The case prediction model was loaded from the local model directory (i.e model.safetensors & spiece.model), and the reasoning was generated based on the query.

 Modifications: Instead of directly copying, we adapted these ideas to fit into the Danswer application by designing a new main.py and modifying Docker configurations, among other changes.

2) (main.pv) Backend:

Path: cd danswer/backend/danswer/caseprediction/main.py:

<u>Summary:</u> The backend for the Case Prediction feature was implemented using FastAPI, which provides a POST endpoint that accepts legal case queries and returns the prediction along with reasoning. The backend also interacts with Hugging Face's models for both prediction and reasoning generation.

Key Modifications:

 Imported Hugging Face model loading, tokenization, and configuration for prediction tasks.

Code:

<u>from transformers import AutoModelForSequenceClassification, AutoTokenizer, AutoConfig</u>

- → AutoModelForSequenceClassification: Loads a pre-trained model for sequence classification (binary classification in this case).
- → AutoTokenizer: Loads the tokenizer corresponding to the model for input tokenization.
- → AutoConfig: Loads the configuration of the model.

Code:

<u>load_dotenv()</u>
 <u>HUGGINGFACE_API_TOKEN = os.getenv("HUGGINGFACEHUB_API_TOKEN")</u>

<u>Purpose:</u> Loads the API token for Hugging Face from the .env file, allowing the backend to authenticate and access Hugging Face models for reasoning generation.

.env: HUGGINGFACEHUB API TOKEN=(hf YwDknPwmCDSrckmFFIfouOEgMeTDPAYBJY)

Code:

- MODEL_PATH = "/app"

 config = AutoConfig.from_pretrained(MODEL_PATH)
 tokenizer = AutoTokenizer.from_pretrained(MODEL_PATH, config=config, local_files_only=True, use_fast=False)
 model = AutoModelForSequenceClassification.from_pretrained(MODEL_PATH, config=config, local_files_only=True)
- → Purpose: Loads the model, tokenizer, and configuration from a pre-defined directory inside the Docker container. These are required to process incoming queries and make predictions using the sequence classification model.

Code:

- <u>Ilm_client = HuggingFaceHub(repo_id="mistralai/Mixtral-8x7B-Instruct-v0.1", model_kwargs={"temperature": 0.2, "max_length": 10000})</u>
- → Purpose: Initializes the Hugging Face LLM (Large Language Model) client using the langchain package. This model is used to generate reasoning based on the case prediction result.
- class CaseQuery(BaseModel):
 query: str

class PredictionResult(BaseModel): prediction: int reasoning: str

- → <u>Purpose:</u> Defines the input and output models using Pydantic's BaseModel. These models structure the API request body and response, ensuring data is validated and returned in a predictable format:
- CaseQuery: Contains the query for case prediction.
- PredictionResult: Contains the prediction (Accepted/Rejected) and reasoning text.

Code:

<u>@app.post("/caseprediction", response_model=PredictionResult)</u>
 <u>async def case_prediction(query_data: CaseQuery):</u>
 <u>try:</u>

Tokenization

```
inputs = tokenizer(query_data.query, return_tensors="pt", padding=True,
truncation=True, max_length=512)
   # Model inference
  with torch.no grad():
   outputs = model(**inputs)
    predictions = torch.argmax(outputs.logits, dim=-1)
      prediction = predictions.numpy()[0]
 # Generate label
 label = "accepted" if prediction == 1 else "rejected"
 # Generate reasoning using LLM
reasoning prompt = f"The following case was {label} based on the analysis:
{query data.query}"
 reasoning = Ilm_client.generate(
 prompts=[reasoning_prompt],
    max new tokens=1000,
  temperature=0.2
 reasoning text = reasoning.generations[0][0].text.strip()
 # Prepare response
response content = {"prediction": int(prediction), "reasoning":
reasoning text}
return jsonable_encoder(response_content)
except Exception as e:
logger.error(f"Error processing case prediction: {str(e)}")
return JSONResponse(status_code=500, content={"detail": str(e)})
```

<u>Purpose:</u> The /caseprediction endpoint processes a legal case query and performs the following steps:

- Tokenization: The input query is tokenized using the pre-loaded Hugging Face tokenizer.
- **Prediction**: The **model predicts** whether the **case** will be **accepted** or **rejected**.
- Reasoning Generation: Using the Hugging Face Hub client, reasoning is generated based on the prediction.
- Response: The prediction and reasoning are returned as a JSON response using jsonable_encoder.

3) Dockerfile for caseprediction setup:

<u>Summary:</u> This **Dockerfile sets** up the **environment** and **dependencies** for the **Case Prediction service**.

- Key Modifications:
- Model File Copying:

Code:

- <u>COPY model.safetensors config.json spiece.model special tokens map.json tokenizer config.json /app/</u>
- → Copies model files (including model.safetensors, config.json, special_toknes_map.json, tokenizer_config.json and tokenizer files) into the container.
- Application Setup: Copies the application files (main.py, .env, and config.py) and sets environment variables.

Code:

COPY main.py ./
COPY config.py ./
COPY .env ./

- Why is MODEL_PATH set to /app?
- → In Docker-based environments, every container runs in its own isolated filesystem.

 When you build a Docker image, you define a working directory inside the container where all the files will be stored and accessed during runtime. In our case, the working directory for our application is /app inside the Docker container.

In the Dockerfile for the caseprediction service, we have this line:

WORKDIR /app

Now, coming to MODEL_PATH, it's set to /app because that's where we're copying all the necessary files, including the model files (model.safetensors, config.json, special_tokens_map.json, tokenizer_config.json and spiece.model) that are required for loading and running the model. This ensures that the tokenizer and

model can find the correct files in the /app directory when they are loaded in the code.

4) (ChatPage.tsx) Frontend:

Path: cd danswer/web/src/app/chat/ChatPage.tsx:

Summary: In the ChatPage.tsx file, several key modifications were made to integrate the Case Analysis Assistant.

- **Key Modifications:**
- <u>API Call Logic Based on Selected Assistant:</u> A new function was added to check if the selected assistant is the Case Analysis Assistant and route the API request accordingly:

Code:

```
// Helper function to check if the selected assistant is Case Analysis
const isCaseAnalysisAssistant = (assistant: Persona | null): boolean => {
  return assistant?.name === "Case Analysis"; // Assuming we know the
  assistant's name
};
```

This function determines if the current assistant selected by the user is the Case Analysis assistant.

 Conditional API Call for Case Analysis Assistant: The onSubmit function was modified to make an API call to the Case Prediction backend when the Case Analysis Assistant is selected

Code:

```
if (isCaseAnalysisAssistant(liveAssistant)) {
   console.log("Case Analysis assistant selected, sending request...");
```

```
headers: {
 "Content-Type": "application/json",
if (!response.ok) {
 console.error(
 "Error with Case Analysis Assistant:",
 await response.text()
 setPopup({
 message: `Error with Case Analysis Assistant: ${await response.text()}`,
 type: "error",
 });
 return;
const result = await response.json();
 console.log("Received case prediction result:", result);
// Handle the response from the case prediction API
setCompleteMessageDetail((prev) => ({
...prev,
messageMap: new Map(prev.messageMap).set(currChatSessionId, {
 messageld: currChatSessionId,
 message: `Prediction: ${result.prediction ? "Accepted" : "Rejected"
 }\n\nReasoning:\n${result.reasoning}`,
 type: "assistant",
 retrievalType: RetrievalType.None,
    files: [], // Empty array or you can add relevant files if needed
 toolCalls: [], // Empty array or any tool calls if needed
parentMessageId: null, // If there's a parent message, set the appropriate ID,
or use null
 }),
}));
} catch (error) {
setPopup({
 message: "Error with Case Analysis Assistant",
 type: "error",
});
```

setChatState("input"); // Stop further processing once case prediction is handled

<u>return;</u> ง

- If the Case Analysis Assistant is selected, a POST request is made to the /caseprediction API (http://54.79.231.211:8006/caseprediction).
- The user query is sent in the request body.
- The response from the API includes a prediction (accepted/rejected) and reasoning text, which is added to the chat.
- The chat UI is updated with the response from the Case Analysis API, and the request processing is stopped after the case prediction is handled.
- Message Reset After Submission: To ensure the input field is cleared right after the message is submitted, the following line was added in the onSubmit function

resetInputBar(); // Call this function right here to clear the input

- This clears the input bar as soon as a message is submitted, improving the user experience by ensuring that the input area is empty when a new message is being processed.
- 1. PREVIOUS CHAT MESSAGE RECOVER YET TO BE IMPLEMENTED COMPLETED
- 2. STREAMING RESPONSE YET TO BE IMPLEMENTTED
- 5) docker-compose.dev.yml

Path: cd danswer/deployment/docker-compose/docker-compose.dev.yml:

<u>Summary:</u> The docker-compose dev.yml file defines the configuration for running the caseprediction service within the Danswer application. Here's a breakdown of the configuration for the caseprediction service

Key Modifications:

Code:

services:

caseprediction:

build:

context: ../../backend/danswer/caseprediction

dockerfile: Dockerfile

env file:

- ../../backend/danswer/caseprediction/.env

ports:

<u>- "8006:8000"</u>

restart: unless-stopped

volumes:

- ../../backend/danswer/caseprediction:/app

- The caseprediction service is responsible for running the Case Analysis Assistant backend.
- It builds the service from the caseprediction directory using the specified Dockerfile.
- The service exposes port 8006 on the host, allowing access to the case prediction API.
- It uses environment variables from the .env file for configurations like API tokens.
- The volumes section ensures that code and model files from the host machine are available inside the Docker container.

6) personas.yaml:

<u>Summary:</u> The Case Analysis Assistant was introduced by defining a new persona in the personas.yaml file. Each persona in the Danswer application is defined with various parameters that determine how the assistant functions, the prompt it uses, and its appearance in the UI. Here's how the Case Analysis assistant was created in the personas.yaml file

Code:

- id: 4

name: "Case Analysis"

description: >

Assistant for predicting case outcomes and providing legal reasoning.

prompts:

- "CaseAnalysis"

num chunks: 0

Ilm relevance filter: false

Ilm filter extraction: false

recency bias: "no decay"

document_sets: []

icon_shape: 123456 icon_color: "#FFA500" display_priority: 4 is_visible: true

- The Case Analysis Assistant was added to the personas.yaml file with a unique ID

 (4), its own name ("Case Analysis"), and a corresponding "CaseAnalysis" prompt defined in prompts.yaml (Currently Commented Out).
- The assistant is designed to predict case outcomes and generate legal reasoning, without interacting with external document sources (Based on prompts.yaml).
- The assistant was made visible in the UI with a specific icon and color scheme, providing users with a tailored experience for case analysis tasks.

7) Deprecation Warnings to be Noted:

> ARC4 Deprecation in cryptography library:

Code:

/usr/local/lib/python3.11/site-packages/pypdf/_crypt_providers/_cryptography.py:3
2: CryptographyDeprecationWarning: ARC4 has been moved to
cryptography.hazmat.decrepit.ciphers.algorithms.ARC4 and will be removed from this module in 48.0.0.

from cryptography.hazmat.primitives.ciphers.algorithms import AES, ARC4

- Description: The ARC4 cipher is deprecated in the cryptography library and will be removed in version 48.0.0. This warning is triggered by the pypdf package, which is using the ARC4 cipher.
- Action for the Future: Ensure that the pypdf library and any other relevant dependencies are updated to versions that no longer rely on ARC4 before cryptography version 48.0.0 is released.
- Coercing Subquery Object in SQLAlchemy:

Code:

/app/danswer/db/llm.py:119: SAWarning: Coercing Subquery object into a select()
for use in IN(); please pass a select() construct explicitly
 LLMProvider UserGroup.user group id.in (user groups subquery) # type:
ignore

/app/danswer/db/persona.py:192: SAWarning: Coercing Subquery object into a select() for use in IN(); please pass a select() construct explicitly

Persona UserGroup.user group id.in (user groups subquery) # type: ignore

- Description: These warnings indicate that SQLAlchemy is automatically coercing subquery objects into select() constructs for use in an IN() clause. While this behavior is currently supported, SQLAlchemy advises explicitly passing a select() construct in future code to avoid reliance on this automatic coercion.
- Action for the Future: Refactor the relevant SQLAlchemy queries in Ilm.py and persona.py to explicitly use select() constructs in place of subquery objects.

Note: Failed to get max tokens for LLM with name mistralai/Mixtral-8x7B-Instruct-v0.1. Defaulting to 4096. This indicates that the system failed to retrieve the maximum token limit for the specific LLM model mistralai/Mixtral-8x7B-Instruct-v0.1. As a result, it defaulted to a token limit of 4096.

To access the Danswer application as an admin, please log in as:

Email: noelshallum@gmail.com

Password: 12345

Case Analysis Assistant: (http://54.79.231.211/chat?assistantId=-4)

- Ibrahim Sultan