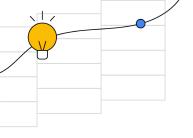


Guidelines

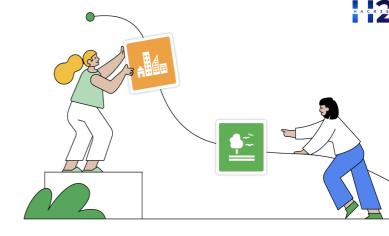
- Kindly use the given template for submitting your project (Make a copy of the template).
- One team is only required to submit one project.
- The ideal size of the presentation should not be more than 10-12 slides.
- You are welcome to add as many POCs and design concepts to support your project.
- It is mandatory to use Gemini APIs while building your prototype.
- In case of queries, kindly reach out to us at apacsolutionchallenge@hack2skill.com













Al for a better tomorrow

Team Details

- a. Team name: onemanarmy
- b. Team leader name: Trần Võ Minh Hùng
- c. Problem Statement: The problem is to design and implement an Al assistant that can effectively understand and process multimodal user inputs (text, voice, images) related to plant health issues, utilize different LLM architectures (local Gemma and cloud Gemini) for analysis and response generation, and maintain conversational context through summarization and RAG to provide users with relevant preliminary diagnoses and actionable recommendations.



Brief about your solution

Solution Name: "Al-Powered Plant Health Assistant,"

Brief: This project is an Al-powered, multimodal chatbot assistant designed to help users identify potential plant health issues. Users can interact with the assistant via text or voice and can upload images of affected plants. The system leverages Large Language Models (LLMs) specifically a local Gemma-3-4B-IT model and the cloud-based Gemini 2.0 Flash API – to understand user gueries, analyze image descriptions, provide preliminary diagnoses, and offer actionable advice. Key features include session management, conversation history, image analysis, voice-to-text, text-to-speech, and a RAG (Retrieval Augmented Generation) system for contextually relevant information retrieval from past conversation summaries and image descriptions. The system architecture allows for dynamic switching between different LLMs for response generation and includes an "evaluation mode" to compare or enhance responses using a more powerful model (Gemini).





Opportunities

How different is it from any of the other existing ideas?

<u>Multimodal & Multi-LLM Flexibility:</u> While many plant disease apps exist, this solution uniquely integrates multiple LLMs (local Gemma and cloud Gemini) and provides a framework to switch between them or use them in tandem (e.g., Gemini for evaluation). This allows for experimentation and leveraging the strengths of different models. The multimodal input (text, voice, image) is comprehensive.

<u>Conversational & Iterative Diagnosis:</u> Unlike static image lookup tools or simple Q&A bots, this system engages in an iterative diagnostic conversation, asking clarifying questions based on initial input (including AI-generated image descriptions) to gather more details, mimicking a consultation with an expert.

<u>Integrated RAG and Summarization:</u> The continuous summarization of the conversation and embedding of key information (including image descriptions) into a vector database for RAG ensures that the LLM has relevant, session-specific long-term memory, leading to more coherent and context-aware interactions over extended conversations.

<u>Developer-Focused Features (Toggles):</u> The inclusion of toggles for switching LLMs and activating an "evaluation mode" makes it a powerful platform for research, development, and comparative analysis of LLM performance in this specific domain, which is not typically a feature of end-user plant apps.





Opportunities

How will it be able to solve the problem?

<u>Accessibility:</u> Provides users (e.g., home gardeners, small-scale farmers) with an accessible first line of support for plant health concerns, available 24/7.

<u>Information Gathering:</u> Systematically guides users to provide relevant information (plant type, symptoms, environment) that is crucial for diagnosis, which users might not initially know to offer. <u>Preliminary Analysis:</u> Offers potential causes for plant issues based on the provided information, helping users narrow down possibilities.

<u>Actionable Advice:</u> Suggests immediate, safe management steps and preventative measures. Reduces Diagnostic Uncertainty: While not a replacement for lab tests, it helps users understand potential issues better and make more informed decisions about plant care or when to seek professional help.





Opportunities

USP of the proposed solution

<u>Adaptive Al Core:</u> The ability to switch and evaluate different LLM backbones (Gemma vs. Gemini) for the core diagnostic task.

<u>Contextual Conversational Diagnosis:</u> Deeply contextual conversations powered by RAG over session summaries and image descriptions.

<u>Multimodal Interaction</u>: Seamless handling of text, voice, and image inputs for a richer user experience.

<u>Developer/Researcher-Friendly:</u> Built-in toggles and evaluation mechanisms facilitate ongoing improvement and experimentation with Al models.





List of features offered by the solution

User Interaction:

- + Text-based chat interface.
- + Voice input (Speech-to-Text via Whisper-small).
- + Text-to-Speech output (via Kokoro TTS).
- + Image upload capability for visual diagnosis.
- + Live camera capture for image input.

Al Capabilities:

- + Primary Response Generation:
- + Using local Gemma-3-4B-IT model.
- + Option to switch to Google's Gemini 2.0 Flash API.
- + Image Description Generation: AI (Gemma, or potentially Gemini if main is Gemini) describes uploaded images to extract visual symptoms.
- + Clarification Question Generation: Al asks targeted questions to gather more diagnostic details.
- + Preliminary Diagnosis & Advice: Al suggests potential plant issues and management strategies.
- + Conversational Context & History Management:
 - ++Stores and retrieves chat session history (SQLite).
 - ++Maintains short-term memory for LLM context.

•





List of features offered by the solution

- +RAG (Retrieval Augmented Generation):
- ++Embeds user messages, Al responses, and image descriptions into a vector database (ChromaDB) per session.
- ++Retrieves relevant past conversation snippets and image context to augment prompts for the LLM.
- + Conversation Summarization:
- ++Periodically summarizes the ongoing conversation using Qwen2.5-0.5B-Instruct with a structured template.
- ++Uses this summary as part of the RAG context.
- + "Eval Mode":
- ++Allows triggering an additional, enhanced response from Gemini based on the current conversation summary and latest image, for comparison or improved insights.
- + Session Management (Frontend & Backend):
- ++Create, rename, delete, and switch between chat sessions.
- ++Persistence of sessions and messages.
- + Backend & System:
- ++WebSocket server for real-time communication.
- ++Asynchronous task handling.
- ++Persistent storage for API keys (via settings.json).
- ++Developer toggles for switching LLM and activating evaluation mode.



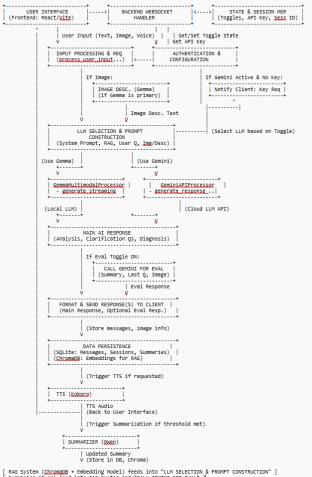


Process flow diagram - a link to view: https://drive.google.com/file/d/1M6jVVndR50O_ sY4-gxjuyC9pPsAa9dl9/view?usp=sharing



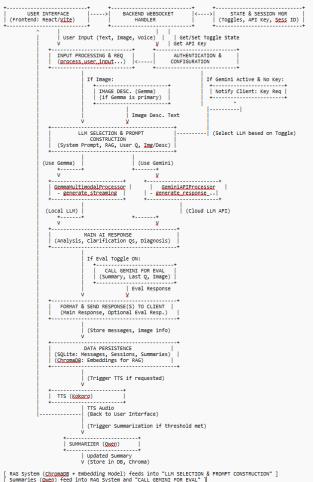


- 1. User Interaction (Frontend):
- + The user interacts with the React frontend (e.g., MainChatView.tsx).
- + They can type a text message, upload an image file, or capture an image using the camera. Voice input is also possible, which is first transcribed to text by Whisper on the backend.
- + The frontend client also manages UI elements for toggles (Gemini, Eval, Grounding) and an API key input mechanism.
- 2. Input Transmission (Frontend to Backend):
- + When the user sends a message (potentially with an image) or changes a toggle state, or submits an API key, the frontend formats a JSON message.
- + This message is sent over a WebSocket connection (managed by WebSocketProvider.tsx) to the Python backend server.
- + Relevant sessionId, message type (e.g., text_input, update_toggle_state, set_api_key), and data (text, image base64, toggle name/state, API key) are included.





- 3. Backend Reception & Initial Handling (handle_client & receive_data_from_client):
- + The backend's handle_client function manages the WebSocket connection for each client.
- + The nested receive_data_from_client task listens for incoming messages.
- + Session Management: The handler identifies or establishes the active_client_session_id.
- + Toggle/API Key State Update: If the message is update_toggle_state or set_api_key:
- ++The backend updates its global state variables (e.g., GLOBAL_USE_GEMINI_MODEL, GEMINI_API_KEY_STORE).
- ++The API key is saved persistently (e.g., to settings.json).
- ++An acknowledgment (..._ack) is sent back to the client. If enabling a Gemini feature without a key, an error/prompt for key is sent.
- ++If an API key is set, the gemini_processor_instance for that client handler is initialized.

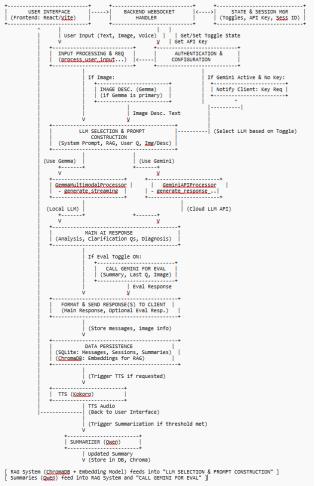








- 4. Main Input Processing (process_user_input_and_respond):
- + For a text_input message:
- ++The user's message ID is generated, and text/image data is extracted.
- ++If an image is present, it's decoded. current_image_bytes_for_gemini and current_image_pil_for_turn (for Gemma) are prepared. The image file is saved to media_storage.
- ++The user's message (text and image filename reference) is saved to the SQLite database (messages table).
- ++Gemma's short-term history is updated with the user's turn (this aids RAG even if Gemini is used later).





Process flow diagram or Use-case diagram

- 5. LLM Selection & Pre-processing:
- + The function checks the GLOBAL_USE_GEMINI_MODEL toggle and if a gemini_processor instance is available (i.e., API key is set and processor initialized).
- + If Gemini Path:
- ++active_llm_service is set to "Gemini".
- ++RAG context (existing conversation summary from Qwen + relevant snippets from ChromaDB) is retrieved via gemma processor. retrieve rag context().
- ++The detailed

GEMINI_CHAT_SYSTEM_PROMPT_FOR_MAIN_RESPONSE is prepared.

- + If Gemma Path (Fallback or Default):
- ++active_llm_service is set to "Gemma".
- ++Gemma Image Description (if image present):

gemma_processor.describe_image() is called. The generated description is sent to the client immediately, saved to the DB, and embedded into ChromaDB. This description (image_description_for_gemma_context) is then available for Gemma's main diagnostic step.

- ++RAG context is retrieved (implicitly within gemma_processor.generate_streaming which calls _retrieve_rag_context).
- ++The REVISED_SIMPLIFIED_SYSTEM_PROMPT is used.

```
rontend: React/Vite)
                               IMAGE DESC. (Gemma)
                                                                     | Notify Client: Key Rea
                              (if Gemma is primary)
                                          Image Desc. Text
                       LLM SELECTION & PROMPT

    (Select LLM based on Toggle)

                (System Prompt, RAG, User Q, Img/Desc)
                GemmaMultimodalProcessor
                                                           (Cloud LLM API)
                (Analysis, Clarification Qs, Diagnosis)
                           If Eval Toggle ON:
                                CALL GEMINI FOR EVAL
                               (Summary, Last Q, Image)
                (SOLite: Messages, Sessions, Summaries)
                (ChromaDB: Embeddings for RAG)
                          (Trigger TTS if requested)
                TTS (Kokono)
                          (Back to User Interface)
                           (Trigger Summarization if threshold met)
                         V (Store in DB, Chroma)
Summaries (Qwen) feed into RAG System and "CALL GEMINI FOR EVAL" ]
```





- 6. Main Al Response Generation:
- + Gemini Path:
- ++gemini_processor.generate_response_for_chat() is called with the system prompt, user query, RAG context, and direct image bytes (if any). It uses Google Search tool if needed.
- + Gemma Path:
- ++gemma_processor.generate_streaming() is called with the user query, (Gemma-generated) image description, and implicitly uses RAG context and its system prompt.
- + The chosen LLM generates the main Al response (e.g., analysis, clarification questions, preliminary diagnosis).
- 7. Post-Main Response Processing & Sending:
- + The generated Al response (ai_final_response_text_for_client) is sanitized.
- + The response is sent to the client via WebSocket.
- + The Al response is saved to the SQLite database (messages table), tagged with the active_llm_service.
- + If Gemma was used, its internal short-term history is updated.
- + TTS Generation (Optional): If requested (generate_tts is true), the Al's text response is sent to tts_processor.synthesize_initial_speech() (Kokoro), and the resulting audio is streamed back to the client.

```
(Frontend: React/Vite)
                                                                    If Gemini Active & No Kev:
                               IMAGE DESC. (Gemma)
                                                                     | Notify Client: Key Rea
                              (if Gemma is primary)
                                          Image Desc. Text
                       LLM SELECTION & PROMPT

    (Select LLM based on Toggle)

                            CONSTRUCTION
                (System Prompt, RAG, User Q, Img/Desc)
                GemmaMultimodalProcessor
             (Local LIM)
                                                            (Cloud LLM API)
                (Analysis, Clarification Qs, Diagnosis)
                           If Eval Toggle ON:
                                CALL GEMINI FOR EVAL
                               (Summary, Last Q, Image)
                (SOLite: Messages, Sessions, Summaries)
                (ChromaDB: Embeddings for RAG)
                           (Trigger TTS if requested)
                TTS (Kokono)
                           TTS Audio
                           (Back to User Interface)
                           (Trigger Summarization if threshold met)
                         V (Store in DB, Chroma)
 RAG System (ChromaDB + Embedding Model) feeds into "LLM SELECTION & PROMPT CONSTRUCTION" ]
Summaries (Qwen) feed into RAG System and "CALL GEMINI FOR EVAL" ]
```



- 8. Evaluation Mode (If Eval Toggle ON):
- + If GLOBAL_EVAL_MODE_ACTIVE is true and gemini_processor is available:
- ++The latest conversation summary (from Qwen, via gemma_processor's session state) is retrieved.
- ++The user's current query (user_text_for_db) and the current turn's image (current_image_bytes_for_gemini) are prepared. gemini_processor.generate_eval_response() is called with the Eval-specific system prompt, summary, query, and image.
- ++The Eval response (prefixed with <EVAL>) is sent to the client as a distinct message (type: "eval_response", sender: "Al_Evaluator").
- ++ The Eval response is saved to the database.
- 9. Conversation Summarization (Background Trigger):
- + After a certain number of message exchanges (SUMMARY_TRIGGER_THRESHOLD), update_session_summary is called as a background task.
- + Recent conversation turns are fetched from the database. GLOBAL_SUMMARIZER.summarize_text() (Qwen-Instruct) is called with the new turns and the existing running summary.
- + The new, updated, structured summary is stored in the session_summaries SQLite table and gemma_processor's session state, and its embedding is upserted into ChromaDB for future RAG.

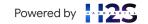
```
rontend: React/Vite)
                               IMAGE DESC. (Gemma)
                              (if Gemma is primary)
                       LLM SELECTION & PROMPT
                                                                    (Select LLM based on Toggle)
                (System Prompt, RAG, User Q, Img/Desc)
                                                            (Cloud LLM API)
                (Analysis, Clarification Qs, Diagnosis)
                           If Eval Toggle ON:
                                 CALL GEMINI FOR EVAL
                               (Summary, Last Q, Image)
                (SOLite: Messages, Sessions, Summaries)
                (ChromaDB: Embeddings for RAG)
                           (Trigger TTS if requested)
                           (Back to User Interface)
Summaries (Qwen) feed into RAG System and "CALL GEMINI FOR EVAL" ]
```





Architecture diagram of the proposed solution - a link to view https://drive.google.com/file/d/1VHYnZHmTXHtP1ClgHvhSr2KucsZ1lWgq/view?usp=sharing





Architecture diagram of the proposed solution

- 1. User Interface (Top Box Frontend):
- + The user interacts here, providing input (text, image uploads, camera captures, voice via microphone).
- + UI components like MainChatView, Sidebar, Toggles, and ApiKeyModal manage these inputs.
- + User actions are translated into WebSocket messages (JSON payloads) by the WebSocketProvider.
- + This layer also receives responses (chat messages, audio) from the backend to display to the user.
- 2. WebSocket Connection (Vertical Line):
- + Represents the persistent, real-time communication channel between the frontend and backend.

```
USER INTERFACE (Frontend)
[Input: Text, Image, Voice] ----> [ MainChatView, Sidebar, Toggles, APIKeyModal ]
       User Actions
                                             WebSocket Messages (JSON)
[Display: Chat, Audio] <----- [ WebSocketProvider (Manages WS & Client State) ]
                                               | WebSocket Connection (ws://...)
    [ WebSocket Server (handle_client) ] <-----
    [ Request Router / Dispatcher (receive_data_from_client) ]
                                             | Global Toggle/API Key States
              process_user_input_and_respond |
                                         (In-memory + settings.json)
                             Main AI Resp
                   (Image) | + Eval Resp.
[AudioSegmentDetector] | [GemmaMultimodalProcessor] | [GeminiAPIProcessor]
                (Local Gemma-3)
                                    (Gemini API)
    [Whisper ASR] +----> [Local Gemma Model]
                                                 [Google Gemini API]
(Speech-to-Text)
                   RAG Context, User Query, Image/Desc.
        +---> [ RAG System ] -------> [ChromaDB (Vectors)]
                   Embeddings +---> [SentenceTransformer]
                +---- [ Conversation Summarizer (Owen) ]
                           Store | Store Summary
    [ SQLite DB (Messages, Sessions, Summaries) ] & [ Media Files (Images) ]
                         AI Resp. Text for TTS
    [ Kokoro TTS ] <----+
          Audio out to client
```





Architecture diagram of the proposed solution

- 3. Backend Server (Bottom Box Python):
- + WebSocket Server (handle_client): The entry point for incoming WebSocket connections and messages.
- + Request Router (receive_data_from_client): Parses messages.
- ++ If it's a settings message (API key, toggle change), it updates the Global Toggle/API Key States. The API key is also persisted to settings.json.
- ++ If it's user input (text, image, or transcribed voice), it's passed to process user input and respond.
- ++ Voice data (from realtime_input) goes to AudioSegmentDetector (VAD).
- + VAD & ASR:
- ++ AudioSegmentDetector identifies speech.
- ++ WhisperTranscriber converts speech to text. This text then flows into process_user_input_and_respond.
- + process user input and respond (Main Logic):
- ++ This is the central orchestrator.
- ++ If an image is provided and Gemma is the primary LLM, it first calls GemmaMultimodalProcessor for an Image Description.
- ++ It then constructs a full prompt including:
- +++A System Prompt (either for Gemma or Gemini).
- +++Context from the RAG System.
- +++The User's Query.
- +++The Image itself (for Gemini) or the Image Description text (for Gemma).
- ++ It selects either GemmaMultimodalProcessor or GeminiAPIProcessor based on the active toggle and API key availability.

 (cont.)

```
[Input: Text, Image, Voice] ----> [ MainChatView, Sidebar, Toggles, APIKeyModal ]
        User Actions
                                               WebSocket Messages (JSON)
[Display: Chat, Audio] <----- [ WebSocketProvider (Manages WS & Client State) ]
                                                 WebSocket Connection (ws://...)
     [ WebSocket Server (handle_client) ] <-----
     [ Request Router / Dispatcher (receive data from client) ]
                                             (In-memory + settings.json)
                    (Image) | + Eval Resp.
[AudioSegmentDetector] | [GemmaMultimodalProcessor] | [GeminiAPIProcessor]
                (Local Gemma-3)
                                      (Gemini API)
     [Whisper ASR] +----> [Local Gemma Model]
                                                  [Google Gemini API]
(Speech-to-Text)
                    RAG Context, User Query, Image/Desc.
         +---> [ RAG System ] ------> [ChromaDB (Vectors)]
                                   +---> [SentenceTransformer]
                +---- [ Conversation Summarizer (Owen) ]
                           Store | Store Summary
     [ SQLite DB (Messages, Sessions, Summaries) ] & [ Media Files (Images) ]
                          AI Resp. Text for TTS
     [ Kokoro TTS ] <----+
```





Architecture diagram of the proposed solution

- 3. Backend Server (Bottom Box Python):
- + LLM Processors:
- ++ GemmaMultimodalProcessor interacts with the Local Gemma Model.
- ++ GeminiAPIProcessor interacts with the Google Gemini API (Cloud).
- + Main Al Response: The selected LLM generates the primary response (analysis, clarification, diagnosis).
- + Eval Mode: If the "Eval" toggle is active, Main Logic makes an additional call to GeminiAPIProcessor (using the latest summary, user query, and image) to get an evaluation response.
- + Format & Send Responses: Main Logic sends the main AI response (and optionally the Eval response) back through the WebSocket Server to the User Interface.
- + Data Persistence: All user messages, Al responses, image filenames, and summaries are stored in the SQLite DB. Image files go to Media Storage. Embeddings for RAG go into ChromaDB.
- + RAG System:
- ++ Uses SentenceTransformer to create embeddings.

Stores/retrieves from ChromaDB.

- ++ Provides context (past snippets, summaries) to the LLM prompt construction phase.
- + Conversation Summarizer (Qwen):
- ++ Periodically takes conversation chunks.
- ++ Generates a structured summary.
- ++ The summary is stored in SQLite and its embedding in ChromaDB (feeding back into RAG).
- + TTS (Kokoro):
- ++ If requested, the Al's text response is converted to speech.
- ++ The audio data is sent back through the WebSocket Server to the User Interface for playback.

```
[Input: Text, Image, Voice] ----> [ MainChatView, Sidebar, Toggles, APIKeyModal ]
        User Actions
                                              WebSocket Messages (JSON)
[Display: Chat, Audio] <----- [ WebSocketProvider (Manages WS & Client State) ]
                                                | WebSocket Connection (ws://...)
     [ WebSocket Server (handle_client) ] <------
     [ Request Router / Dispatcher (receive data from client) ]
              process_user_input_and_respond |
                                          (In-memory + settings.json)
                             Main AI Resp.
                   (Image) | + Eval Resp.
[AudioSegmentDetector] | [GemmaMultimodalProcessor] | [GeminiAPIProcessor]
                (Local Gemma-3)
                                    (Gemini API)
    [Whisper ASR] +----> [Local Gemma Model]
                                                 [Google Gemini API]
(Speech-to-Text)
                   RAG Context, User Query, Image/Desc.
         +---> [ RAG System ] ------> [ChromaDB (Vectors)]
                   Embeddings +---> [SentenceTransformer]
                +---- [ Conversation Summarizer (Owen) ]
                           Store | Store Summary
     [ SQLite DB (Messages, Sessions, Summaries) ] & [ Media Files (Images) ]
                          AI Resp. Text for TTS
    [ Kokoro TTS ] <----+
          Audio out to client
```

Architecture diagram of the proposed solution

Backend Server (Bottom Box - Python):

WebSocket Server (handle_client): The entry point for incoming WebSocket connections and messages.

Request Router (receive_data_from_client):

Parses messages.

If it's a settings message (API key, toggle change), it updates the Global Toggle/API Key States. The API key is also persisted to settings.json.

If it's user input (text, image, or transcribed voice), it's passed to process_user_input_and_respond.

Voice data (from realtime_input) goes to

AudioSegmentDetector (VAD).

VAD & ASR:

AudioSegmentDetector identifies speech.

WhisperTranscriber converts speech to text. This text then flows into

process_user_input_and_respond.

process_user_input_and_respond (Main Logic):

This is the central orchestrator.

If an image is provided and Gemma is the primary

```
[Input: Text, Image, Voice] ----> [ MainChatView, Sidebar, Toggles, APIKeyModal ]
        User Actions
                                               WebSocket Messages (JSON)
[Display: Chat, Audio] <----- [ WebSocketProvider (Manages WS & Client State) ]
                                                  WebSocket Connection (ws://...)
     [ WebSocket Server (handle_client) ] <-----
     [ Request Router / Dispatcher (receive data from client) ]
                    (Image) | + Eval Resp.
[AudioSegmentDetector] | [GemmaMultimodalProcessor] | [GeminiAPIProcessor]
                                                   [Google Gemini API]
(Speech-to-Text)
         +---> [ RAG System ] ------> [ChromaDB (Vectors)]
                                    +---> [SentenceTransformer]
                +---- [ Conversation Summarizer (Owen) ]
                            Store | Store Summary
     [ SQLite DB (Messages, Sessions, Summaries) ] & [ Media Files (Images) ]
                            AI Resp. Text for TTS
     [ Kokoro TTS ] <----+
```







Technologies to be used in the solution

(Mandatory to integrate Gemini APIs)

I. Frontend (Client-Side)

Core Framework/Library:

- + React: For building the user interface components.
- + Vite: As the build tool and development server (providing fast HMR).
- + TypeScript: For static typing, improving code quality and maintainability.

State Management & Hooks:

- + React Hooks (useState, useEffect, useCallback, useMemo, useRef): For managing component state and lifecycle.
- + React Context API (createContext, useContext): Used in WebSocketProvider and useWebSocket for managing and providing WebSocket connection and related state globally to components.

Communication:

Native Browser WebSocket API: For real-time, bidirectional communication with the backend.

UI Components & Styling:

- + Shadcn/ui (Implied/Assumed): Based on components like <Button>, <Input>, <ScrollArea>, <Card>, <Switch>, <Label>, <Dialog>, this is a likely candidate for pre-built, accessible UI components. If not Shadcn/ui, then another similar component library or custom-built components with Tailwind CSS.
- + Tailwind CSS (Implied by tailwind.config.js): For utility-first CSS styling.
- + lucide-react: For icons.
- + react-markdown: For rendering Al responses that might contain Markdown formatting.

Utilities:

- + js-base64: For Base64 encoding/decoding (e.g., image data, audio data).
- + uuid: For generating unique identifiers (e.g., for messages, sessions).

Audio Processing (In-Browser):

+ Web Audio API (AudioContext, MediaStreamAudioSourceNode, AudioWorkletNode): For capturing and processing microphone audio for VAD/ASR. audio-processor.js (Custom AudioWorklet): For handling audio buffering and potentially downsampling/formatting before sending to the backend.





Technologies to be used in the solution

- II. Backend (Server-Side)
- 1. Core Language & Framework:
- + Python 3.10+: The primary programming language.
- + AsynclO: For concurrent programming, handling WebSocket connections and asynchronous LLM/API calls.
- 2. WebSocket Server:
- + websockets library: For creating and managing the WebSocket server endpoint.
- 3. Machine Learning / Al Models & Libraries:
- + Hugging Face transformers library: For loading and interacting with various pre-trained models.
- + Hugging Face sentence-transformers library: For generating text embeddings for RAG.
- + PyTorch: As the underlying deep learning framework for the Hugging Face models.
- + bitsandbytes: For 4-bit quantization of LLMs (like Gemma and Whisper) to reduce VRAM usage.







Technologies to be used in the solution

- 4. Models Used:
- + Gemma-3-4B-IT (unsloth/gemma-3-4b-it-qat-bnb-4bit): Local multimodal LLM for primary/fallback responses and image descriptions.
- + Whisper-small (openai/whisper-small): For Automatic Speech Recognition (ASR).
- + Kokoro-82M (hexgrad/Kokoro-82M): For Text-to-Speech (TTS).
- + Qwen2.5-0.5B-Instruct (unsloth/Qwen2.5-0.5B-Instruct-bnb-4bit or similar): For conversation summarization.
- all-MiniLM-L6-v2 (sentence-transformers/all-MiniLM-L6-v2): For generating embeddings for RAG.
- + Google google-genai library: For interacting with the Google Gemini API.
- + Gemini 2.0 Flash API (gemini-2.0-flash): Cloud-based LLM for primary responses (if toggled) and evaluation responses. Includes Google Search tool integration.
- 5. Data Storage & Retrieval:
- + SQLite3 (sqlite3 module): For relational database storage of chat sessions, messages (including metadata like image filenames, data types, LLM used), and full conversation summaries.
- + ChromaDB (chromadb client): Persistent vector database for storing and querying embeddings of conversation turns, image descriptions, and summaries to facilitate RAG.
- + File System (os, shutil modules):
- ++ For storing uploaded/captured image files (media_storage/ directory).
- ++ For persisting the Gemini API key (server_settings.json).

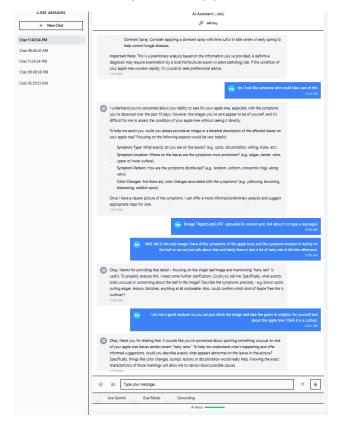
6. Utilities:

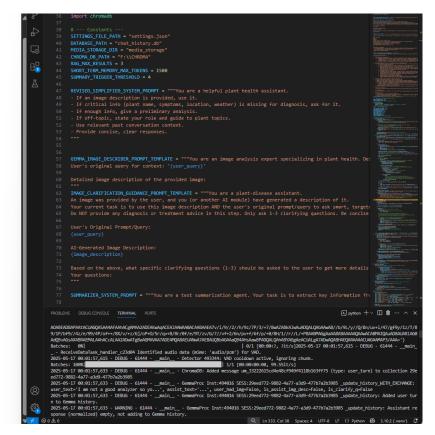
- + json module: For serializing/deserializing JSON messages.
- + base64 module: For decoding image/audio data from the client.
- + Pillow (PIL): For image manipulation (e.g., opening image bytes for Gemma).
- + numpy: For numerical operations, especially with audio data and embeddings.
- + logging module: For server-side logging.
- + re module: For regular expressions (e.g., in strip_markdown_for_tts).
- + uuid module: For generating unique IDs on the backend.
- + functools.partial: Used for wrapping calls for run_in_executor.





Snapshots of the prototype



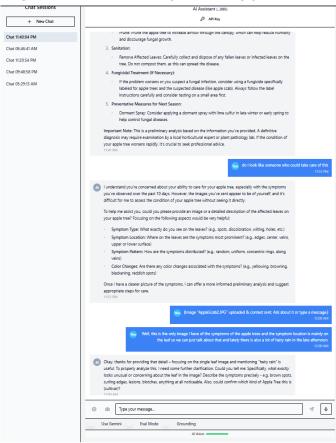


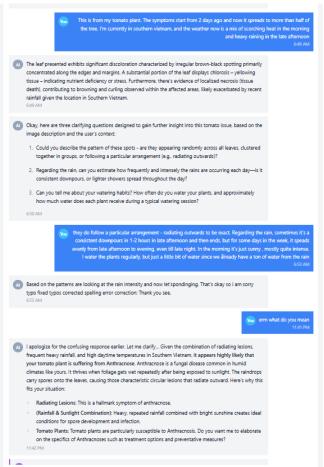






Snapshots of the prototype





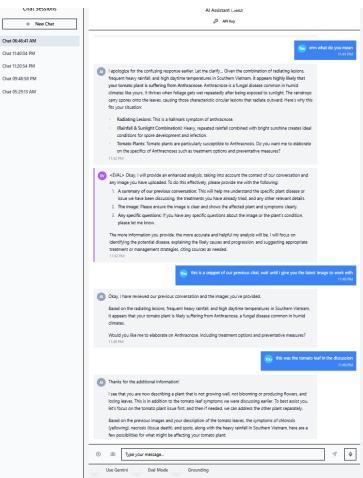








Snapshots of the prototype

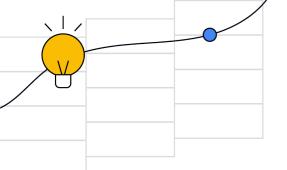






Additional Details/Future Development (if any):

- 1. Grounding with my custom dataset:
- https://github.com/alberttrann/plant-disease-dataset_and_finetuning_geminihttps://huggingface.co/datasets/minhhungg/plant-disease-dataset
- Tittps://tiuggingrace.co/datasets/friii initialigg/plant-disea
- 2. Grounding with user's custom "Knowledge Base"









Google Developer Groups

Solution Challenge





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