# **Documentation**

# Documentation: Weather & Wine Recommendation System

#### **Overview**

This project fetches weather and wine data, merges them, stores them in a PostgreSQL database, uses OpenAI's GPT to generate wine recommendations based on current weather, and exposes the functionality through a FastAPI backend.

#### **Modules Breakdown**

#### wine\_fetch.py - Fetch Wine Descriptions

- Purpose: Uses the Spoonacular API to fetch wine descriptions for a predefined list of wines.
- **Why**: Descriptive metadata about each wine helps the language model make more context-aware recommendations.
- Key Features:
  - Asynchronous API calls using aiohttp.
  - Saves data to Wine\_train.json

#### data\_fetch.py - Weather Data Collection

- **Purpose**: Uses the OpenWeather API to fetch **current weather data** for a predefined list of cities.
- Why: Weather conditions (e.g. temperature, feels-like temperature) influence wine preferences (e.g. bold reds vs. light whites). Supplying weather data helps the language model make contextual and personalized wine recommendations.

#### Key Features:

- Asynchronous weather data fetching using aiohttp for efficient parallel API calls.
- Extracts relevant metrics such as temperature and feels-like temperature in both Celsius and Fahrenheit.
- Saves:
  - Raw weather data to Weather\_train.json
  - Cleaned and structured data to weather\_cleaned.json

#### merge\_data.py - Merge Weather and Wine Data

- Purpose: Merges weather data and wine descriptions into a single JSON file (merged\_data.json).
- Why: Centralizing the data allows the LLM module to access and reason over a unified structure.

#### Key Features:

- Asynchronous file I/O using aiofiles.
- Structure ensures the merged output includes keys "weather" and "wine" for easy parsing downstream.

#### Database.py - Store Merged Data in PostgreSQL

- Purpose: Inserts the content of merged\_data.json into a PostgreSQL table (merged\_data).
- **Why**: Persistent storage ensures data availability across sessions and scales better than keeping everything in memory.

#### Key Features:

- Auto-creates table if it doesn't exist.
- Stores entire JSON blob for flexibility in future querying or auditing.

#### Ilm.py - Generate LLM-Based Wine Recommendations

- **Purpose**: Uses OpenAI's GPT to generate summaries that recommend a wine based on a user's weather-based query.
- **Why**: The power of LLMs allows for rich, contextual wine recommendations by interpreting temperature and wine features.

#### Key Features:

- Extracts the city from the query.
- Looks up the weather for that city.
- Constructs a natural-language prompt to GPT-4.
- Stores unique summaries in the analysis\_summaries table in PostgreSQL.
- Avoids duplicate entries using a pre-check.

#### main.py - FastAPI Web Server

- **Purpose**: Hosts three API endpoints to interact with the system.
- Why: Provides an interface for external clients to use the system programmatically.

#### • Endpoints:

- POST /fetch\_and\_process: Accepts a query like "What's the weather in Paris and what wine suits it?" Calls the LLM and stores result.
- GET /results: Returns all LLM-generated recommendations, optionally filtered by city.
- GET /analysis: Returns only the latest summary (or latest per city).

#### **Data Flow Overview**

1. wine\_fetch.py + (external weather data source): Generate base JSONs.

- 2. merge\_data.py: Combines both datasets into merged\_data.json.
- 3. Database.py: Saves this merged data to the PostgreSQL database.
- 4. Im.py: Reads from the merged file, calls GPT, and stores summaries.
- 5. main.py: API interface to query the system and access summaries.

## **Files Summary**

File	Role
wine_fetch.py	Fetch wine data from Spoonacular API
merge_data.py	Merge wine + weather JSON into unified format
merged_data.json	Output file from merge step (used throughout the system)
Database.py	Push merged data into PostgreSQL
llm.py	GPT logic to generate and store recommendations
main.py	FastAPI backend for querying and managing data
.env	Secure storage of API keys and DB credentials (not shared here)

# Why Use GPT/LLM?

- The use of GPT enables **natural language interpretation** and **contextual wine pairing** that would be difficult to hardcode.
- By combining weather details (e.g., "feels like 34°C") with nuanced wine descriptions, GPT provides thoughtful and dynamic recommendations.

#### **Environment Variables**

The system relies on a **.env** file with:

POSTGRES\_DB\_URL=your\_postgres\_connection\_url

# OPENAI\_API\_KEY=your\_openai\_api\_key SPOONACULAR\_API\_KEY=your\_spoonacular\_api\_key

```
Kindly use own api-keys

Open-Weather = ""

spoonacular= " "

Open_API_key = ""

PostgressSQL = ""
```

# Why This Architecture?

- **Separation of concerns**: Each script handles one task (data fetching, merging, storing, generating, serving).
- **Scalability**: Modular design allows replacing components (e.g., new wine API or weather provider).
- **Efficiency**: Async operations for fetching and merging improve performance.
- Reliability: PostgreSQL provides durable storage, while LLM ensures rich content generation.

#### **LLM Integration Overview**

This is a visual representation of how the **LLM (Large Language Model)** integrates the weather and wine data:

- The weather API provides city-specific temperature and climate data.
- The wine API offers descriptive profiles of various wines.
- When a user asks, for example, "What wine should I have in Paris?", the system:
  - 1. Looks up the current weather in Paris.
  - 2. Uses GPT to interpret both the **weather** and the **wine dataset**.
  - 3. Generates a personalized wine recommendation.

#### **Adaptive Recommendations**

The output varies based on city and weather. For example:

- In Paris with warm weather, it may suggest a light, crisp white wine like Riesling.
- In **Berlin** on a cooler day, it might recommend a bold red like **Cabernet Sauvignon**.

### **Python Testing**

- Catch Bugs Early You can find issues before they make it to production (like broken routes, DB errors, or incorrect logic).
- Ensure Code Quality Tests enforce a contract—if someone changes the logic in main.py, your tests can immediately catch regressions.
- Enable Confident Refactoring Want to change something in your code? Run the tests afterward to confirm nothing broke.
- Automate Validation Manually checking each endpoint or function is tedious.
   With pytest, you can validate all endpoints with a single command

#### Illustration

This diagram shows the flow of data and logic between the components:

Freshold Output Debug Control | Member | None
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