Weather-Based Plant Recommendation Engine

1. Function Overview

Goal: This function aims to build an intelligent plant recommendation system. It comprises two
core components: first, a backend data processing workflow that transforms, cleans, and
consolidates scattered raw plant JSON data into a structured, queryable "plant attribute
database"; second, a real-time recommendation logic that can filter and suggest the most suitable
plants from the database based on the 16-day weather forecast for a user-specified geographic
location.

Implementation Approach:

- This is a Rule-Based Expert System, not a machine learning model.
- Core Data Engineering: A key step in this function is converting the abstract agricultural concept of a "Hardiness Zone" into a concrete, comparable metric: the "absolute minimum survival temperature (°C)".
- Core Matching Logic: The "brain" of the recommendation engine is a series of preset conditional rules. It rigorously compares various metrics from the weather forecast (such as minimum temperature, sunshine duration, precipitation, etc.) against the survival requirements of each plant in the database to make a recommendation.

2. Prerequisites & Environment

Libraries:

- Data Processing: pandas, numpy, json, glob
- Data Acquisition: requests
- Database: mysql-connector-python
- Command-Line Tools: argparse (used in the .py script)

• Data Dependencies:

- Local Raw Data: A large number of individual plant JSON files located in the
 01_raw_data/01_species_details directory.
- External API: Open-Meteo Forecast API, used to retrieve 16-day weather forecasts.
- Database: An accessible MySQL database for storing the processed core plant attributes table (Table13 GeneralPlantListforRecommendation).

3. Data Source & Input

Source 1: Raw Plant Data

- **Source**: A collection of scattered plant_species_details_*.json files, with data sourced from the Perenual Plant Open API.
- Data Dictionary (Key Fields):

Field Path	Туре	Description
id	int	Unique ID of the plant
hardiness.min	string	The lowest hardiness zone the plant can tolerate (e.g., "7")
type	string	Plant type (e.g., "tree", "perennial")
sunlight	list	Sunlight requirements (e.g., ["full sun", "part shade"])
watering	string	Watering needs (e.g., "Average", "Minimal")
drought_tolerant	boolean	Whether the plant is drought-tolerant

Source 2: Real-time Weather Forecast

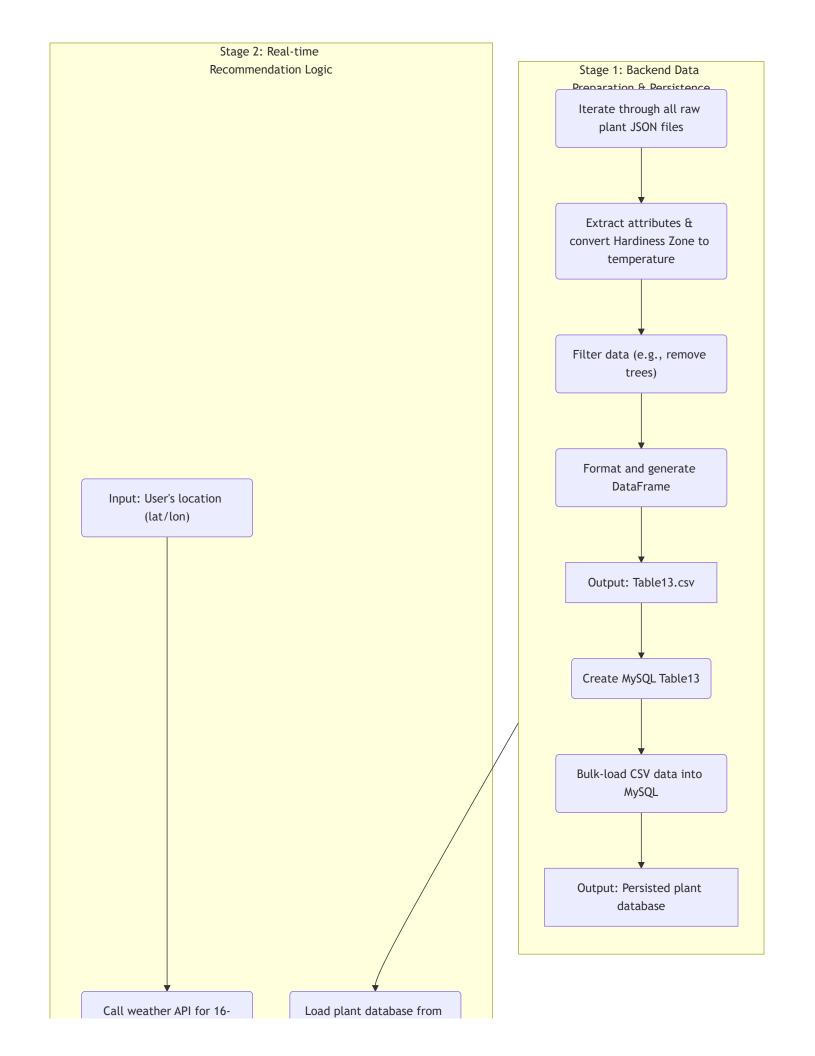
Source: Open-Meteo Forecast API.

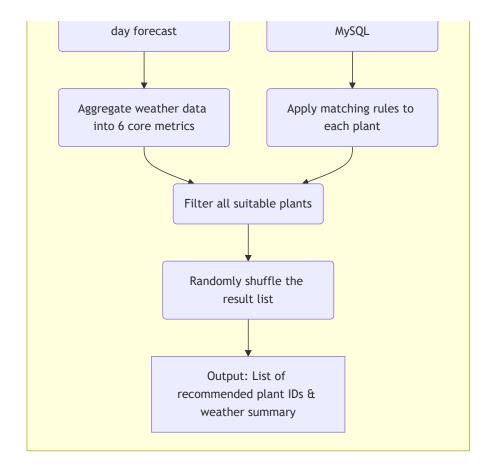
• Input: latitude, longitude.

 Output (Key Daily Variables): temperature_2m_min, sunshine_duration, uv_index_max, precipitation_sum, relative_humidity_2m_mean, etc.

4. Workflow

The workflow for this function is clearly divided into two stages: backend data preparation and persistence, and front-end real-time recommendation logic.





• 4.1. Stage 1: Backend Data Preparation & Persistence

- Step 1: Data Extraction and Transformation: Iterate through all JSON files in the 01_species_details directory. For each file, extract the required plant attributes. The most critical transformation occurs here: using the predefined HARDINESS_ZONE_TO_CELSIUS dictionary, the plant's minimum hardiness zone (e.g., "7") is directly mapped to its absolute minimum survival temperature of -17.8 °C. Concurrently, list data such as sunlight is converted into a JSON string for database storage.
- Step 2: Data Filtering and Formatting: Consolidate all processed plant records into a single Pandas DataFrame. Then, perform filtering, for example, by removing all records where plant_type is 'tree', and organize the column order.
- Step 3: Data Storage (CSV & MySQL): First, save the cleaned and organized DataFrame as a CSV file (Table13_GeneralPlantListforRecommendation.csv) to serve as a backup and intermediate artifact. Subsequently, create the corresponding Table13... table in MySQL and use the LOAD DATA LOCAL INFILE command to efficiently bulk-load the data from the CSV into the database, completing the data persistence process.

• 4.2. Stage 2: Real-time Recommendation Logic

 Step 1: Load Plant Data: When the recommendation engine starts, it first loads the entire contents of Table13 from the MySQL database into a Pandas DataFrame and performs pre-

- processing (such as parsing the sunlight JSON string back into a Python list and converting 0/1 values to booleans).
- Step 2: Fetch and Aggregate Weather Forecast: Receive the latitude and longitude passed from the front end. Call the get_and_aggregate_weather_data function to request a 16-day daily weather forecast from the Open-Meteo API. This 16-day data is then aggregated into 6 core matching metrics: extreme_min_temp, extreme_max_temp, avg_sunshine_duration (in hours), avg_max_uv_index, avg_daily_precipitation, and avg_relative_humidity.
- Step 3: Execute Matching and Filtering: This is the "brain" of the engine. The
 is_plant_suitable function is called, iterating through every row in the plant DataFrame and
 determining if the plant is suitable for the upcoming 16-day weather conditions based on the
 four core matching rules detailed in the next section.
- Step 4: Process and Output Results: Collect the IDs of all plants that pass the filter into a list. To enhance the diversity of the user experience, random.shuffle() is used to randomly shuffle this list. Finally, the aggregated weather information and the randomly sorted list of recommended plant IDs are returned as the result. The .py script version also supports outputting directly in a machine-readable JSON format via the --json command-line argument.

5. Core Algorithm & Rule Interpretation

Algorithm 1: Hardiness Zone to Temperature Conversion

• This is the most central data engineering step in the project. The HARDINESS_ZONE_TO_CELSIUS dictionary is based on the official standard from the U.S. Department of Agriculture (USDA). It acts as a "translator," accurately converting a relatively abstract horticultural "zone" concept into a concrete, computable "temperature" threshold in physics. This step is the foundation upon which all subsequent temperature-related matching rules are built.

Algorithm 2: Recommendation Matching Rules

The is_plant_suitable function contains four cascading rules that form the logical core of the recommendation decision:

- i. Temperature Survival Rule: This is the strictest hard-filter condition. The forecasted extreme minimum temperature (extreme_min_temp) over the next 16 days must be greater than or equal to the plant's absolute minimum survival temperature (absolute_min_temp_c) listed in the database. Any plant that does not meet this condition is immediately eliminated.
- ii. **Sunlight Matching Rule:** This rule considers both the **average sunshine duration** (avg_sunshine_duration) and the **average maximum UV index** (avg_max_uv_index).

- Duration: full sun requires >= 6 hours; part shade requires 3-6 hours; full shade requires < 3 hours.
- Intensity: Additional optimization logic is applied, such as excluding plants that only tolerate part shade from environments with a very high UV index (>8), or excluding fullsun plants from environments with a very low UV index (♥).
- iii. Watering Matching Rule: This rule matches the average daily precipitation

 (avg_daily_precipitation) with the plant's watering needs. For instance, a plant requiring

 Frequent watering would be eliminated if the forecast predicts low daily precipitation (<

 3mm), while a plant needing Minimal watering would be eliminated if the forecast is too wet

 (> 5mm).
- iv. **Drought Tolerance Rule:** This is a supplementary rule. For a plant that is **not** drought-tolerant (drought_tolerant is False), if the **average relative humidity** (avg_relative_humidity) is below 40%, the air is considered too dry, and the plant is deemed unsuitable.

6. Output

• Intermediate Artifacts:

 Table13_GeneralPlantListforRecommendation.csv: The cleaned, transformed, and filtered plant attribute data file.

Final Artifacts:

- The Table13_GeneralPlantListforRecommendation table in the MySQL database, serving as the persisted data source for the recommendation engine.
- For a single successful recommendation request, the final output is a JSON object or formatted text containing two parts:
 - a. **Aggregated Weather**: A summary of the 6 core metrics from the 16-day weather forecast.
 - b. **Recommended Plant IDs**: A randomly shuffled list of plant IDs that satisfy all matching rules.

7. Notes & Future Improvements

Notes:

 The accuracy of the recommendation results is heavily dependent on the accuracy of the Open-Meteo 16-day weather forecast. The 01_plant_recommendation.py script is designed as a standalone service module that
accepts input via command-line arguments and supports JSON output, making it easy to
integrate with other programs (like a backend server).

• Future Improvements:

- Add More Matching Dimensions: More weather variables (e.g., the impact of wind speed on fragile plants) or soil data could be introduced to make the matching rules more refined.
- Make Rules Configurable: The current matching thresholds (e.g., "6 hours for full sun") are hard-coded. They could be designed as configurable parameters to adapt to different scenarios or for A/B testing.
- Introduce a Scoring System: The current system is a binary "suitable/unsuitable" filter. A
 scoring system could be designed to rate all suitable plants based on how well they match,
 providing users with more personalized "highly recommended" vs. "moderately
 recommended" tiers.
- Incorporate Historical Climate Data: In addition to the short-term forecast, local historical climate data could be incorporated to provide a longer-term assessment of climate suitability.