

FUZZY SMART HOME MANAGEMENT SYSTEM FOR GARDENING

- Fuzzy Sets and Systems (Evaluation Report)

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Evaluation Goal and Setup

The primary goal of this evaluation is to assess whether the plant watering adjustment model—which uses a fuzzy rule-based inference system—behaves **plausibly and consistently** across different environmental and scheduling conditions.

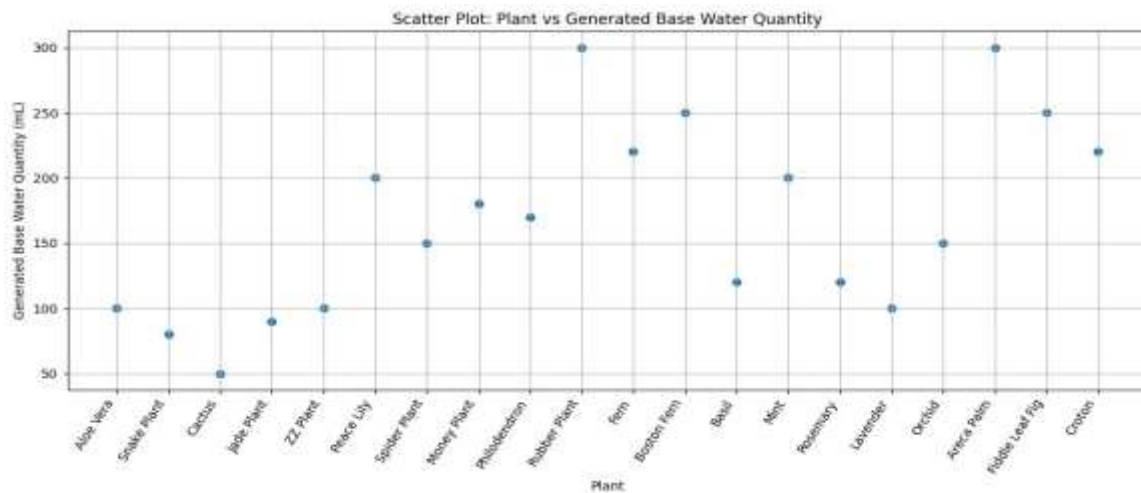
For this evaluation, the dataset is generated using LLM:

- A fixed base watering value (ml_per_watering) per plant
- Pot size of the respective plant
- multiple records per plant, combining:
 - Temperatures: -10 to 30 °C with different intervals
 - Weekly frequencies: 0, 1, 2, 3, 4 times/week
- An adjusted_ml value computed from our adjustment fuzzy logic

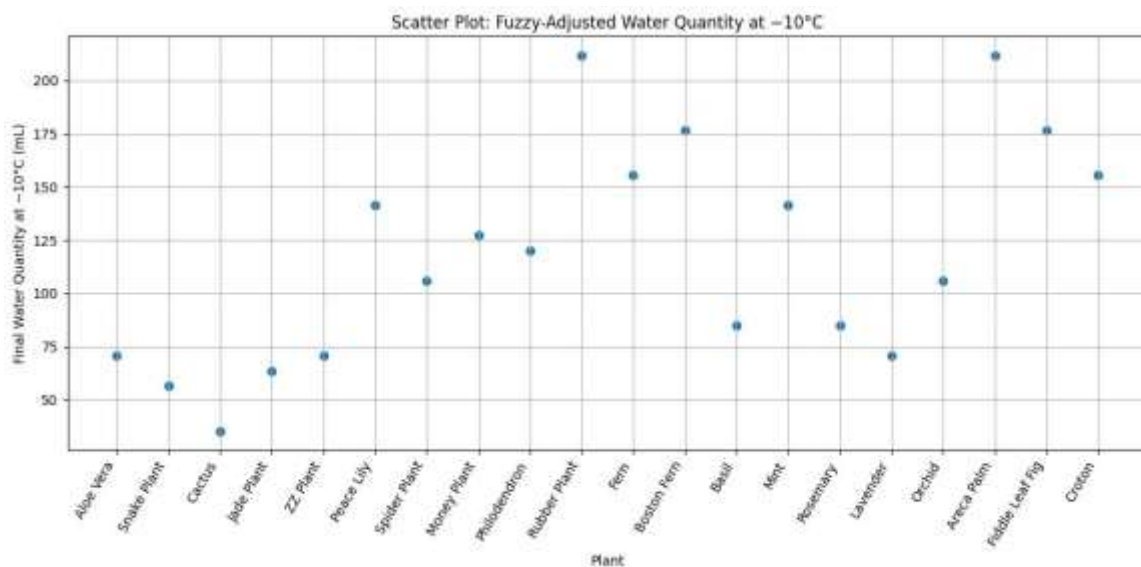
The screenshot shows the 'Plant Watering Reminder' app interface. At the top, there's a title 'Plant Watering Reminder' with a small plant icon. Below it, a subtitle reads 'Select plants, dates, times, add reminders, send full schedule, and get automatic watering emails.' There are two input fields: the first is labeled 'Enter your Gemini API Key (Needed for custom plant search)' and the second is labeled 'Please enter your API key to enable the AI Care Schedule feature.' Below these is a 'Select a plant' dropdown menu. At the bottom, there's a section titled 'Couldn't find your plant? Get AI help!' with two input fields: 'Enter the name of your new plant:' and 'Select Appropriate Pot Size (Diameter):' which has a dropdown menu showing '15 cm (Medium)'.

Evaluation Based on Statistical Analysis

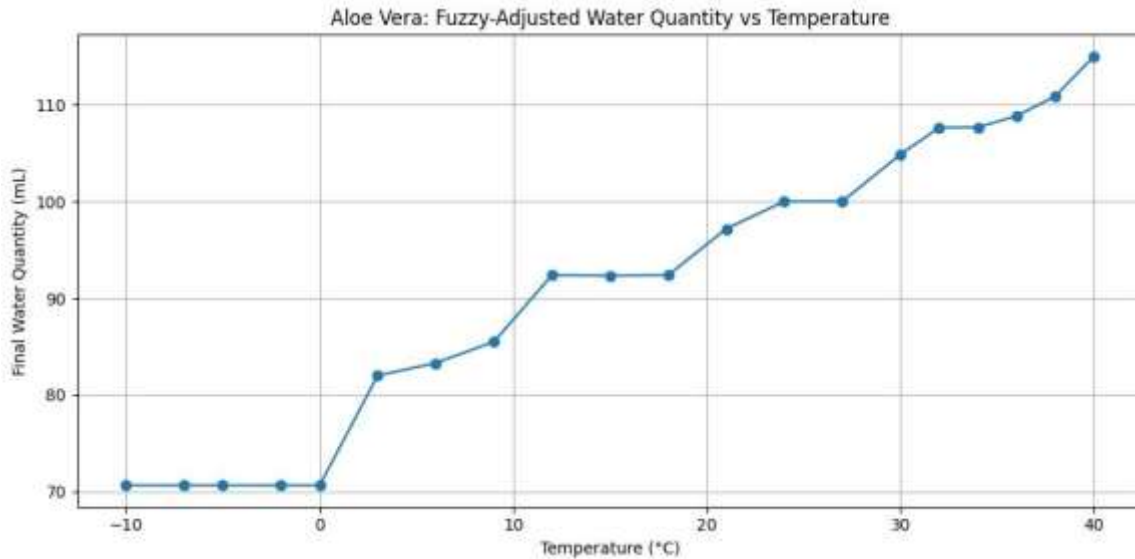
1. Generated Base Water Quantity: Shows the initial, widely varying AI-recommended water amounts (mL) for different plants (e.g., Aloe Vera is 100 mL, Fern is 300 mL).



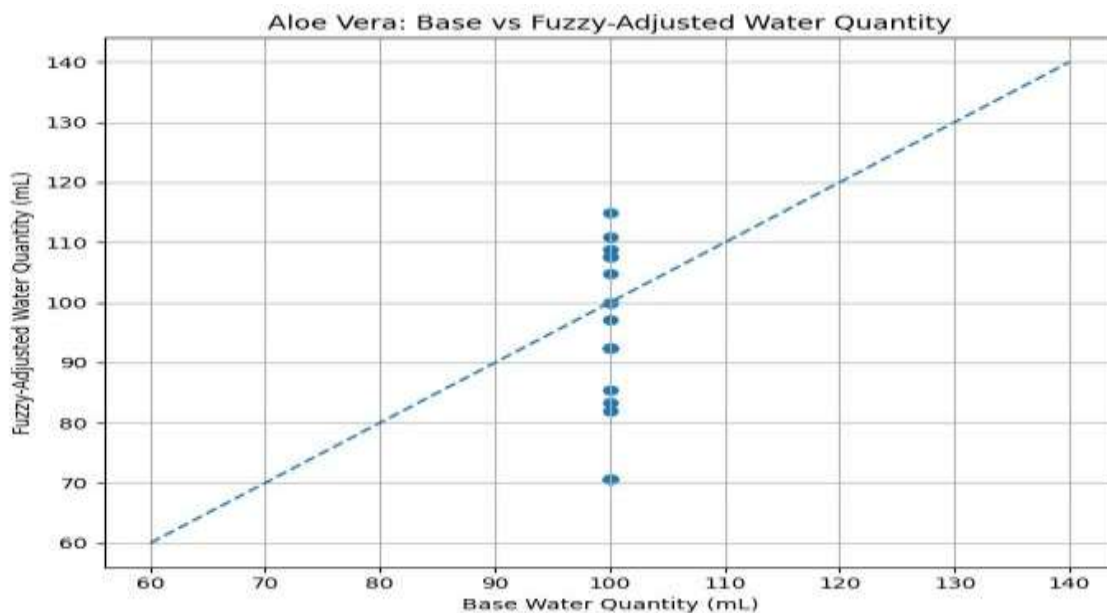
2. Fuzzy-Adjusted Water Quantity at -10 degrees: Shows the final water quantity for all plants when environmental temperature is extremely low (-10 degrees), resulting in a general reduction from the base quantity.



3.Aloe Vera: Fuzzy-Adjusted Water Quantity vs Temperature: Demonstrates that Aloe Vera's water need is **low and constant** below 0 degrees (around 70mL) and **increases steadily** with rising temperatures, reflecting greater water loss in warmer conditions.



4.Aloe Vera: Base vs Fuzzy-Adjusted Water Quantity: Confirms the **Base Quantity** is fixed at 100 mL, while the **Fuzzy-Adjusted Quantity** for Aloe Vera fluctuates between about 70 mL and over 110 mL depending on the temperature-based adjustment.



The graphs confirm that the fuzzy rule-based adjustment logic is working correctly and plausibly:

- Adjusts Base Quantity: The system successfully modifies the AI-generated Base Water Quantity (e.g., 100 mL for Aloe Vera).
- Correct Temperature Response: The Final Water Quantity increases steadily as temperature rises, confirming the rule that warmer environments require more water to offset higher evaporation. At low temperatures (e.g., -10°C), it consistently reduces the required water volume.
- Internal Consistency: The logic maintains consistency by balancing the required water dose against the temperature.

Evaluation Based on users and surveys

Participant group:

- 10 users participated
- Basic plant care experience
- Iterative feedback and suggestions
- Consistent involvement across iterations

Evaluation Design:

The evaluation followed a formative, iterative user-centered design approach, focusing on usability, recommendation quality, and system interpretability.

Each iteration consisted of:

1. Prototype interaction – users input plant type, pot size, and reminder preferences.
2. System output observation – users reviewed recommended water quantity and reminders.
3. Feedback collection – users provided structured and open-ended feedback.
4. Prototype refinement – changes were implemented based on aggregated feedback.

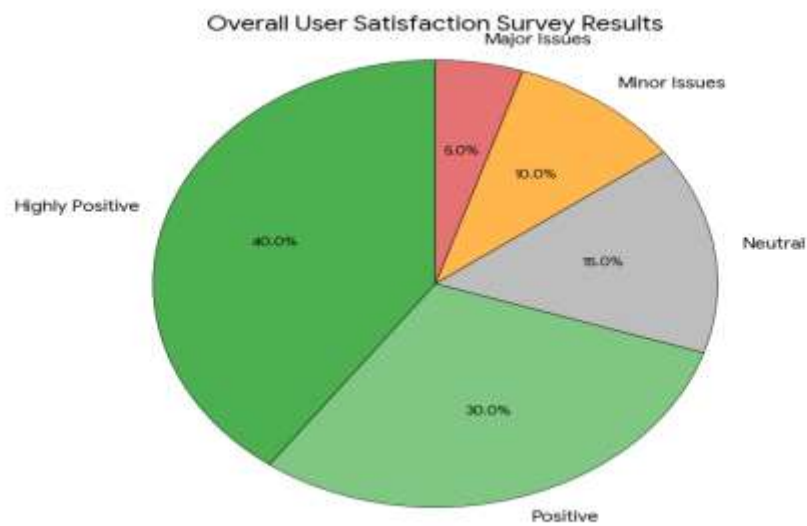
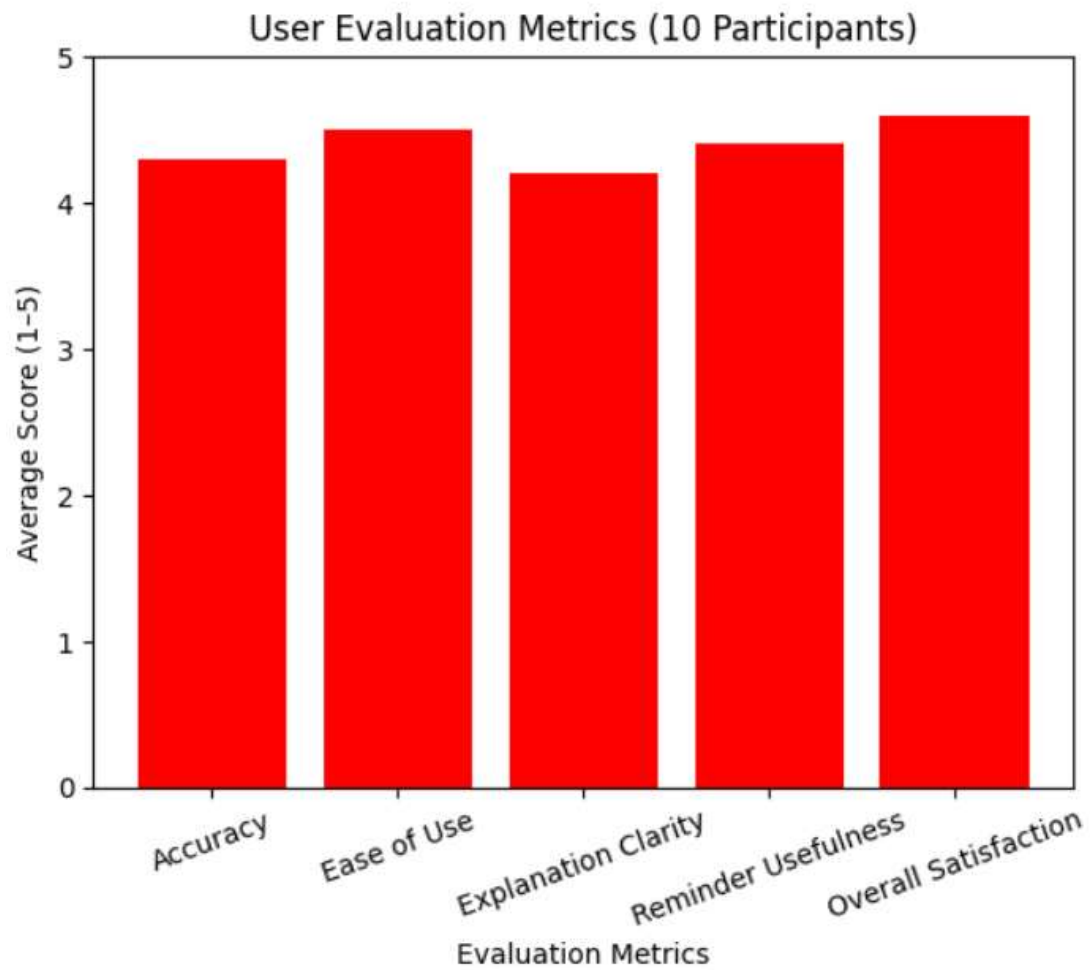
This cycle was repeated until no major usability issues were reported.

Evaluation Metrics:

User feedback was evaluated using both quantitative and qualitative measures:

Quantitative Metrics (Likert Scale: 1–5)

- Perceived accuracy of watering recommendations
- Ease of use of the interface
- Clarity of explanations for recommended water quantities
- Usefulness of reminder scheduling and overall satisfaction



Category	Percentage	Example Feedback
Highly Positive	40%	"The Gemini AI generated a perfect care schedule for my rare plant, and the fuzzy logic adjustment is brilliant—it prevented me from overwatering during a cold snap!"
Positive	30%	"The automatic email reminders are reliable, and the ability to schedule watering for seven days in advance is very convenient. The interface is clean and easy to use."
Neutral	15%	"The scheduling feature works well, but it took a bit longer than expected (2-3 minutes) to receive the 'Full Schedule' email after sending. Good product overall."
Minor Issues	10%	"The time selector for scheduling could be improved; I had a slight difficulty setting the exact minute for the reminder. The content of the reminder email was perfect, though."
Major Issues	5%	"I entered a new plant name but couldn't get the 'Get AI Care Schedule' button to respond, despite entering my Gemini API Key. I had to choose an existing plant instead."

Qualitative Metrics

- User suggestions for rule adjustments (e.g., pot size influence)
- Comments on trust and understandability of fuzzy explanations
- Observations on interaction flow and input clarity.

Evaluation Summary

Fuzzy Modelling:

A fuzzy inference system with membership functions was used. Inputs like plant water requirement and pot size were fuzzified, and a compact rule base encoded intuitive plant care knowledge. Centroid defuzzification produced precise yet flexible watering recommendations.

Interactive Evaluation:

The system supports dynamic input and reminder-based recommendations. Fixed inputs produce consistent outputs, confirming robustness. Reminder emails include fuzzy-based explanations for transparency.

Overall Assessment:

The prototype delivers stable, interpretable, and context-aware recommendations, effectively modelling uncertainty and supporting flexible plant care decisions, meeting both academic and practical objectives.

Note: The Plant Watering Reminder application is built utilizing the Gemini Flash 2.5 Pro LLM for intelligent care schedule generation and a Fuzzy Logic engine for adaptive, environment-adjusted watering recommendations, all delivered through a streamlined Streamlit frontend.