Irrigation Control System Based On Fuzzy Logic

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I. INTRODUCTION

With the current technological advancements, Internet of Things (IoT) has been widely used in the field of machine learning and automation, especially when combined with Artificial Intelligence (AI) technologies. With the help of sensors, data collected can be sent to the server for statistical information and processing [1]. One of the most popular application is smart irrigation systems, which play a huge role in tackling the global water crisis by reducing excess water use. It ensures optimal use of water, and at the same time increases production rate of crops.

In this report, a fuzzy logic-based irrigation system is proposed to determine the length of time that the system must run to apply enough water to meet the plant watering requirement. The fuzzy control system has three input variables and one output variables. The input variables include soil moisture (%), air humidity(%) and temperature(°C), meanwhile the output variables is the duration of valve opening (minutes). The advantage of using a fuzzy logic is its ability to handle uncertainty in lexical semantics. It works according to interval values, such as very long, long, medium, short and very short, as in human logic [2].

II. SYSTEM DESIGN

The fuzzy logic irrigation control system consists of three state (input) variables and one control (output) variable. The unit, value range and linguistic variables of the inputs and outputs are shown in Table I. The controller controls the amount of water to be watered by limiting the duration of the valve opening time.

Firstly, a set of fuzzy rules will be determined. Then, the input will be made fuzzy using the input membership functions as shown in Table II, Table III, Table IV and Table V. All five output membership functions are combined and finally a deffuzzified output distribution is obtained. An output value representing the duration of the valve opening time will be determined based on the input values.

Table I STATE AND CONTROL VARIABLES

Type	Variables	Range	Linguistic Variable
State	Soil Moisture	0-100%	L, M, H
	Air Humidity	0-100%	L, M, H
	Temperature	-10-50°C	L, M, H
Control	Duration	0-10min	VS, S, M, L, VL

A. Membership Functions

The membership functions for each state variable and control variable are provided in this section. The range are determined based on various studies on smart irrigation system [1] [2] [3] [4].

Table II MEMBERSHIP FUNCTIONS OF SOIL MOISTURE

Name of MF	Membership Function	MF Parameters
Low (L)	Trapezoid-Shaped	[0, 0, 20, 40]
Medium (M)	Triangle	[30, 50, 70]
High (H)	Trapezoid-Shaped	[60, 80, 100, 100]

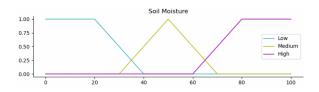


Figure 1. Membership Functions For Soil Moisture

Table III MEMBERSHIP FUNCTIONS OF AIR HUMIDITY

Name of MF	Membership Function	MF Parameters
Low (L)	Trapezoid-Shaped	[0, 0, 15, 40]
Medium (M)	Triangle	[25, 50, 75]
High (H)	Trapezoid-Shaped	[60, 85, 100, 100]

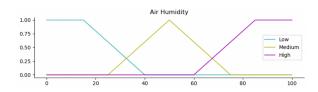


Figure 2. Membership Functions For Air Humidity

Table IV
MEMBERSHIP FUNCTIONS OF TEMPERATURE

Name of MF	Membership Function	MF Parameters
Low (L)	Trapezoid-Shaped	[-10, -10, 0, 15]
Medium (M)	Triangle	[10, 20, 30]
High (H)	Trapezoid-Shaped	[25, 40, 50, 50]

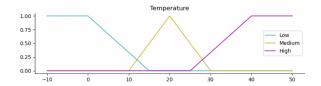


Figure 3. Membership Functions For Temperature

Table V MEMBERSHIP FUNCTIONS OF DURATION

Name of MF	Membership Function	MF Parameters
Very Short (Vs)	Triangle	[0, 0, 2.5]
Short (S)	Triangle	[0, 2.5, 5]
Medium (M)	Triangle	[2.5, 5, 7.5]
Long (L)	Triangle	[5, 7.5, 10]
Very Long (VL)	Triangle	[7.5, 10, 10]

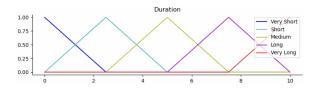


Figure 4. Membership Functions For Duration

B. Fuzzy Rules

Soil moisture, air humidity and temperature have three fuzzy variables each (Low (L), Medium (M) and High (H)), whereas the output variable, duration, has five fuzzy variables (very short (VS), short (S), medium (M), long (L), very long (VL)).

Table VI RULES OF DURATION (IN MIN) AT LOW TEMPERATURE

Soil Moisture	Air Humidity		
	Low	Medium	High
Low	VL	M	S
Medium	M	M	S
High	S	S	VS

1) Rules of duration (in min) at low temperature:

Table VII
RULES OF DURATION (IN MIN) AT MEDIUM TEMPERATURE

Soil Moisture	Air Humidity		
	Low	Medium	High
Low	VL	L	M
Medium	L	M	S
High	M	S	VS

2) Rules of duration (in min) at medium temperature:

Soil Moisture	Air Humidity		
	Low	Medium	High
Low	VL	L	L
Medium	L	M	M
High	L	M	VS

3) Rules of duration (in min) at high temperature:

C. Defuzzification

Defuzzification is achieved by applying the center of gravity method(centroid) from the fuzzification results to obtain the center of area under the curve. This method is widely used for fuzzification as it facilitates smooth transition of defuzzied values around output fuzzy region [5].

III. EXAMPLE OUTPUT

A. Output 1

Values for input and output variables:

Soil moisture: 20% Air Humidity: 15% Temperature: -5°C Duration: 9.17 minutes

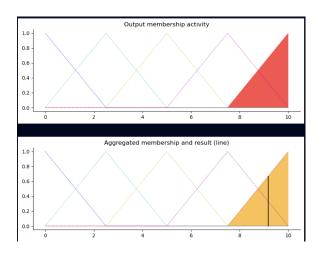


Figure 5. Output 1

B. Output 2

Values for input and output variables:

Soil moisture: 40% Air Humidity: 3% Temperature: 5°C Duration: 5 minutes

C. Output 3

Values for input and output variables:

Soil moisture: 40% Air Humidity: 3% Temperature: 40°C Duration: 7.5 minutes

D. Output 4

Values for input and output variables:

Soil moisture: 80% Air Humidity: 66% Temperature: 40°C Duration: 3.97 minutes

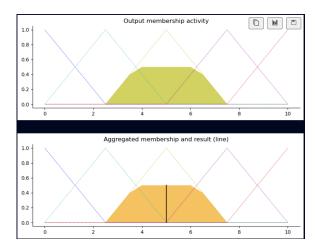


Figure 6. Output 2

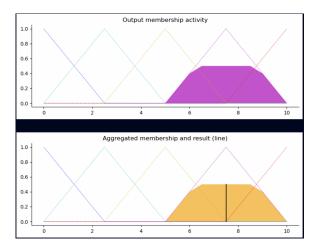


Figure 7. Output 3

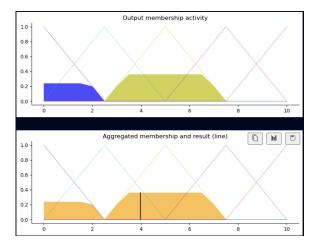


Figure 8. Output 4

E. Output 5

Values for input and output variables:

Soil moisture: 50% Air Humidity: 24% Temperature: 23°C Duration: 7.5 minutes

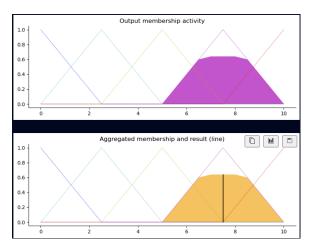


Figure 9. Output 4

IV. DISCUSSIONS

The advantages of applying Fuzzy inferences system in irrigation is the control system ensure optimal water usage by monitoring the soil moisture level, air humidity and temperature. No mathematical model is able to include all input parameters, hence, fuzzy logic technique is used since the membership functions are distributed based on the possible range of each state variable after fuzzification [4]. By using fuzzy set theory, real life uncertainties can be handled effectively, which makes it an ideal choice for smart irrigation control system.

The limitation of this proposed model is that the fuzzy control system is only effective for a single type of crop. This is because the ideal range of the input variables is set according to the crop species.

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