AIDS LAB EXPERIMENT 08

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Aim: To design fuzzy control system using Fuzzy tool /library.

Theory:

A **fuzzy control system** is a control method based on **fuzzy logic**, which allows handling uncertain or imprecise information by using degrees of truth rather than binary true/false values. Unlike traditional controllers, fuzzy controllers use **fuzzy sets** and **linguistic rules** to model complex or nonlinear systems where exact mathematical models are difficult to obtain.

The design process includes:

- Fuzzification: Converting crisp inputs into fuzzy values.
- Rule Evaluation: Applying a set of IF-THEN rules to infer fuzzy outputs.
- **Defuzzification:** Transforming fuzzy outputs into crisp control signals.

Using a **fuzzy tool/library** simplifies this process by providing functions to create membership functions, define rules, and simulate the controller, enabling efficient design and testing of fuzzy control systems.

Designing a Fuzzy Controller

Steps include:

- Selection of input and output variables.
- Defining membership functions for inputs and outputs.
- Constructing fuzzy rules based on expert knowledge or system behavior.
- Implementation and simulation using a fuzzy tool/library.
- Tuning and validation to ensure desired control performance

Advantages of Fuzzy Control Systems

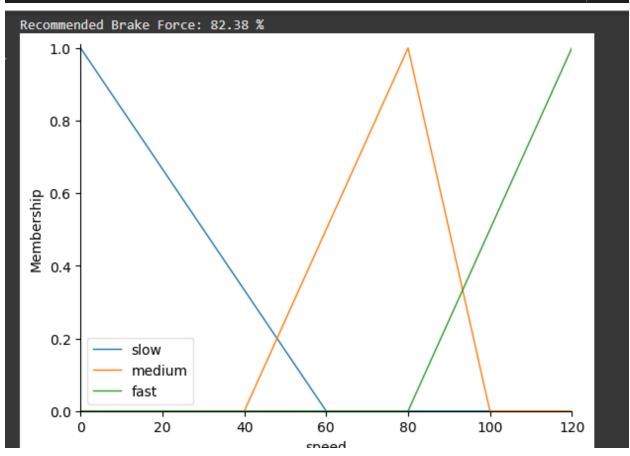
- Handles nonlinear and complex systems effectively
- Robust to noise and uncertainties.
- Incorporates human expert knowledge through linguistic rules.
- No precise mathematical model required.

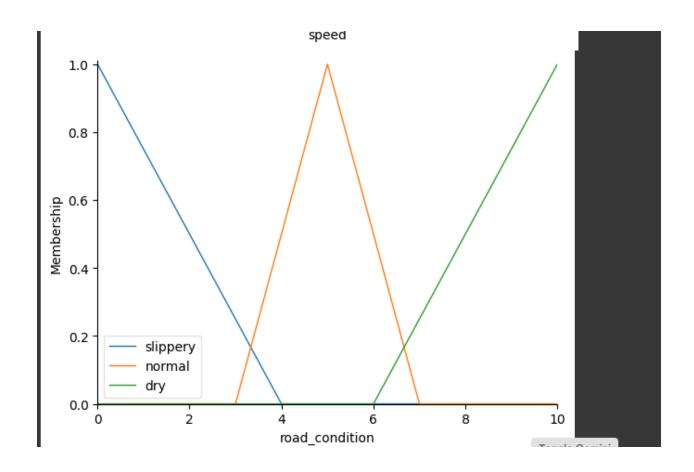
Code:

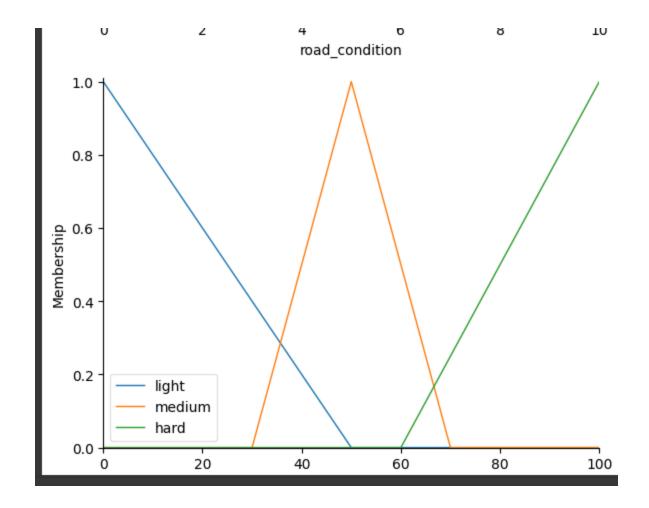
```
[1] !pip install scikit-fuzzy
    !pip install networkx==2.6.3
→ Requirement already satisfied: scikit-fuzzy in /usr/local/lib/python3.11/dist-packages (0.5.0)
    Requirement already satisfied: networkx==2.6.3 in /usr/local/lib/python3.11/dist-packages (2.6.3)
import numpy as np
    import skfuzzy as fuzz
    from skfuzzy import control as ctrl
  speed = ctrl.Antecedent(np.arange(0, 121, 1), 'speed')
  road_condition = ctrl.Antecedent(np.arange(0, 11, 1), 'road_condition') # 0-10
  brake_force = ctrl.Consequent(np.arange(0, 101, 1), 'brake_force') # %
  # --- Membership Functions ---
  # Speed
  speed['slow'] = fuzz.trimf(speed.universe, [0, 0, 60])
  speed['medium'] = fuzz.trimf(speed.universe, [40, 80, 100])
  speed['fast'] = fuzz.trimf(speed.universe, [80, 120, 120])
  # Road Condition (higher is better grip)
  road_condition['slippery'] = fuzz.trimf(road_condition.universe, [0, 0, 4])
  road condition['normal'] = fuzz.trimf(road condition.universe, [3, 5, 7])
  road_condition['dry'] = fuzz.trimf(road_condition.universe, [6, 10, 10])
  # Brake Force
  brake_force['light'] = fuzz.trimf(brake_force.universe, [0, 0, 50])
```

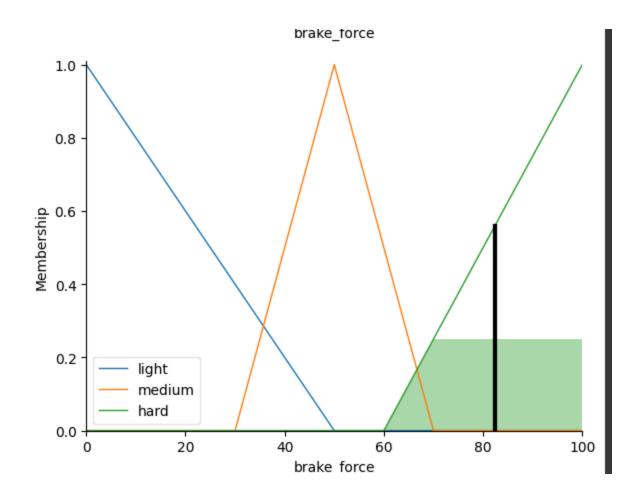
brake_force['medium'] = fuzz.trimf(brake_force.universe, [30, 50, 70])
brake_force['hard'] = fuzz.trimf(brake_force.universe, [60, 100, 100])

```
# --- Define Rules ---
rule1 = ctrl.Rule(speed['fast'] & road_condition['slippery'], brake_force['hard'])
rule2 = ctrl.Rule(speed['fast'] & road_condition['dry'], brake_force['medium'])
rule3 = ctrl.Rule(speed['slow'] & road_condition['dry'], brake_force['light'])
rule4 = ctrl.Rule(speed['medium'] & road_condition['normal'], brake_force['medium'])
rule5 = ctrl.Rule(speed['medium'] & road_condition['slippery'], brake_force['hard'])
rule6 = ctrl.Rule(speed['slow'] & road_condition['slippery'], brake_force['medium'])
# --- Control System ---
brake_ctrl = ctrl.ControlSystem([rule1, rule2, rule3, rule4, rule5, rule6])
brake_sim = ctrl.ControlSystemSimulation(brake_ctrl)
# --- Example Input ---
brake sim.input['speed'] = 90
                                        # km/h
brake_sim.input['road_condition'] = 3 # slippery
# --- Compute Output ---
brake_sim.compute()
print("Recommended Brake Force:", round(brake sim.output['brake force'], 2), "%"|)
```









Conclusion:

The experiment shows that fuzzy logic can effectively model real-world decisions, providing adaptive and smooth control for complex systems.