

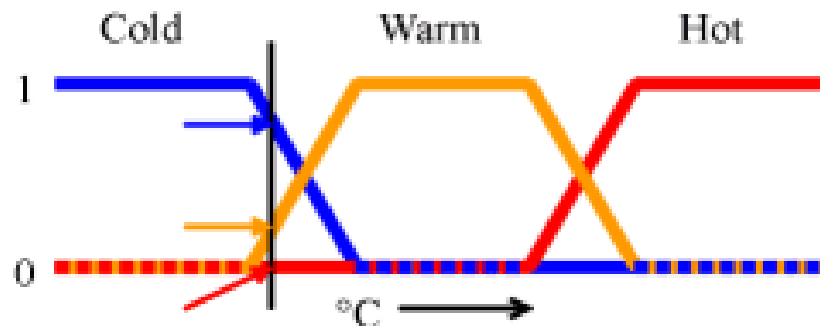
An Introduction to Fuzzy Logic  
2007, Kristian Guillaumier

# Introduction

- Based on fuzzy set theory: true, false, 'in between' – approximate reasoning.
- Concept of a degree of truth: truth is defined as membership in some vaguely defined set (not probability based – no element of randomness).
- Wikipedia example:
  - Bob is in a house with two adjacent rooms: the kitchen and the dining room. In many cases, Bob's status within the set of things "in the kitchen" is completely plain: he's either "in the kitchen" or "not in the kitchen". What about when Bob stands in the doorway? He may be considered "partially in the kitchen". Quantifying this partial state yields a fuzzy set membership. With only his big toe in the dining room, we might say Bob is 99% "in the kitchen" and 1% "in the dining room", for instance. No event (like a coin toss) will resolve Bob to being completely "in the kitchen" or "not in the kitchen", as long as he's standing in that doorway.

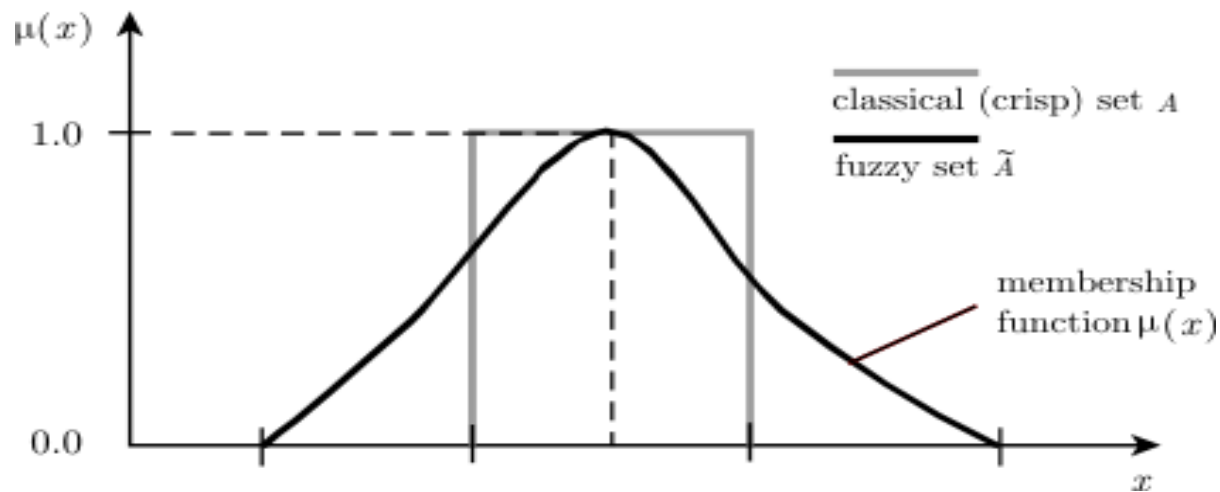
# Set Membership

- Membership values range from 1..0.
  - As opposed to Boolean, 'Crisp' logic.
- This caters for instances of 'sort of', 'almost', 'very', etc...
- Introduced by Lotfi Zadeh (1965).
- Temperature example from Wikipedia:
  - Functions that define cold, warm and hot.
  - Each membership function maps temperatures to a value 1..0.



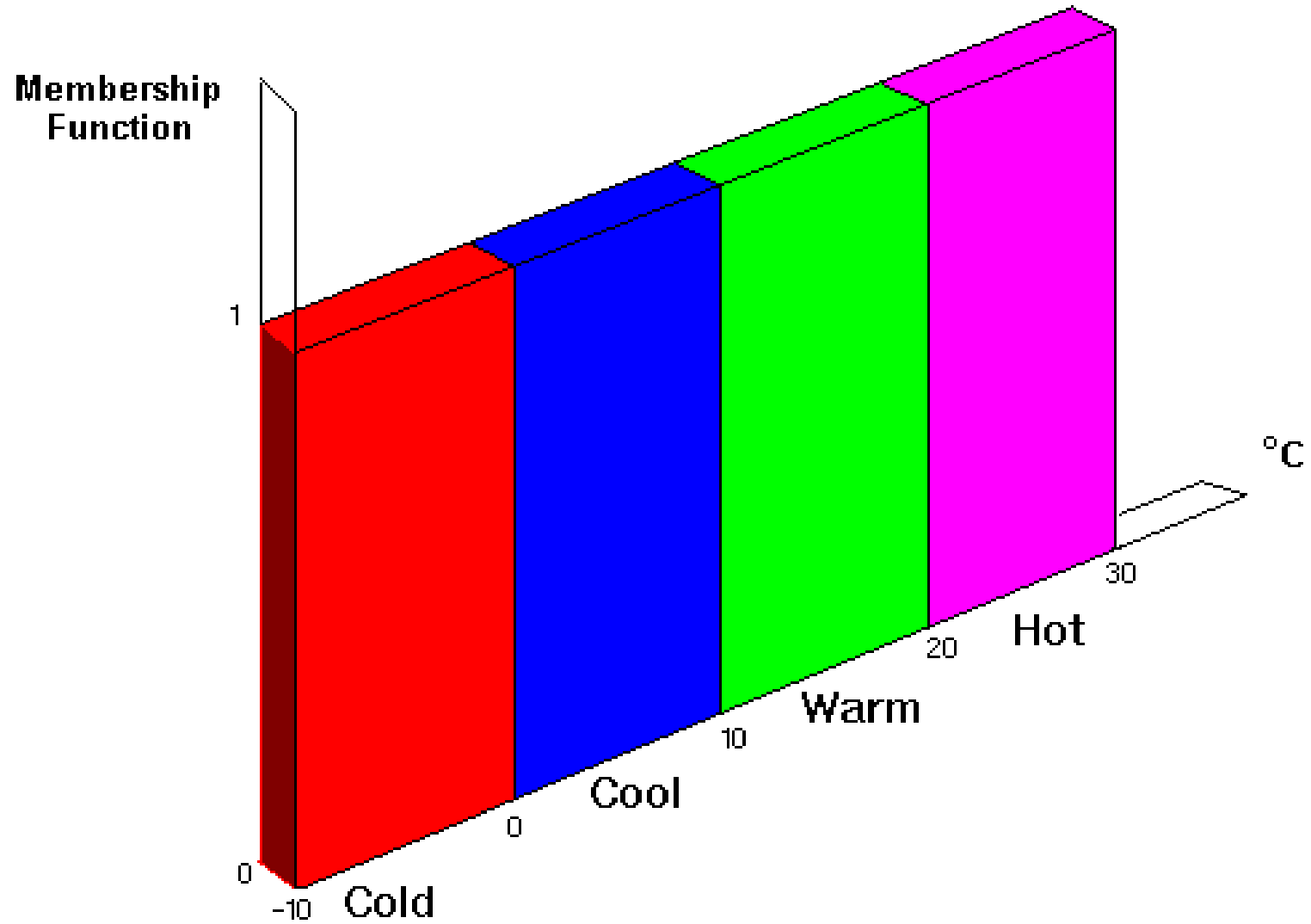
# Set Membership

- Based on a membership function  $\mu \rightarrow [0,1]$ .
- $\mu$  is standard nomenclature for a fuzzy truth value.
- Image from Wikipedia:



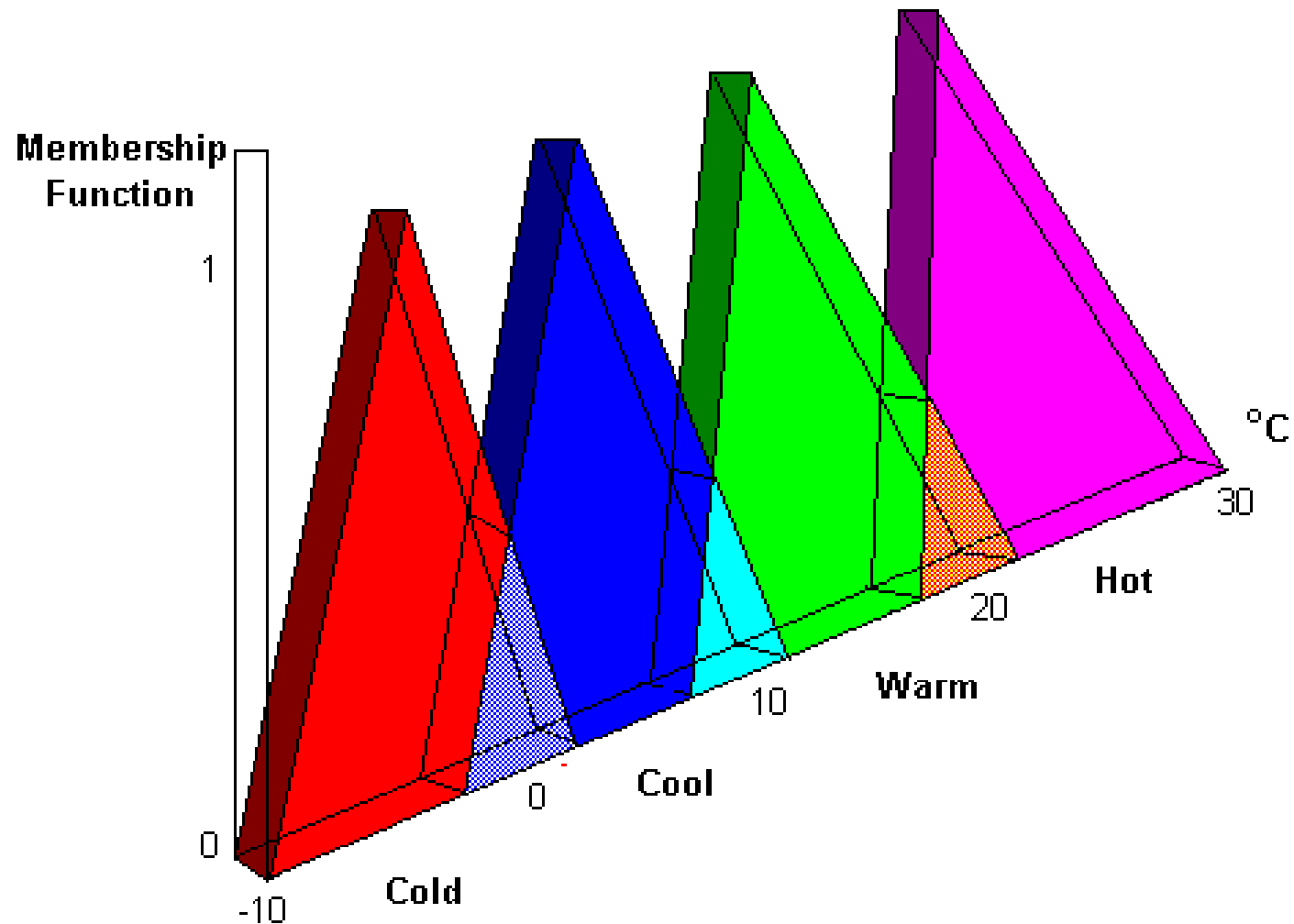
# Set Membership

from [http://www.doc.ic.ac.uk/~nd/surprise\\_96/journal/vol2/jp6/article2.html](http://www.doc.ic.ac.uk/~nd/surprise_96/journal/vol2/jp6/article2.html)



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# Set Membership

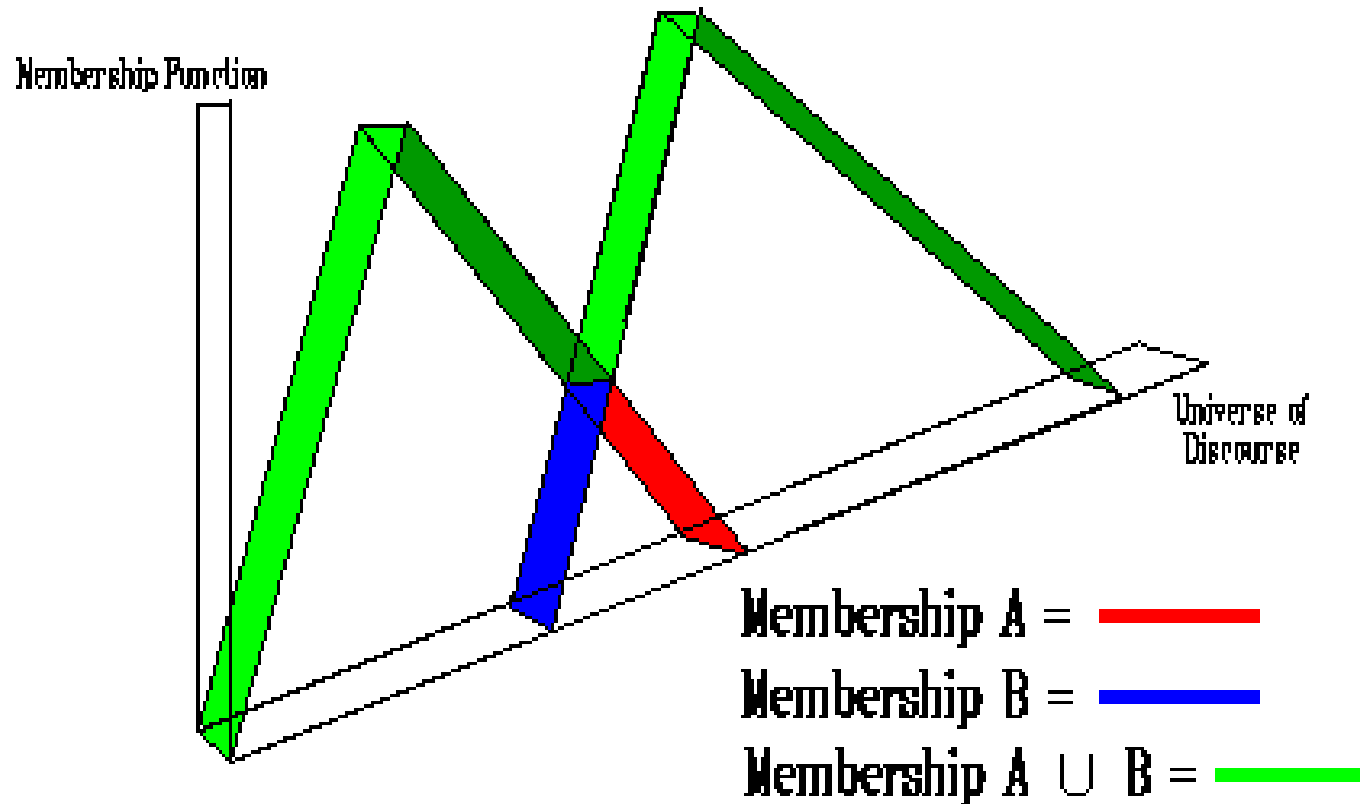
- Based on IF..THEN rules.
- Note there is no ELSE part:
  - Something can be both slightly warm and slightly cold at the same time.
- Membership functions can be triangular, bell-shaped, trapezoidal...

# An Example Membership Function

- $\text{IsTall}(x) =$ 
  - 0;  $x \leq 1.5$ .
  - 1;  $x \geq 2.0$ .
  - $(x - 1.5) / (2.0 - 1.5)$ ;  $x > 1.5, x < 2.0$ .

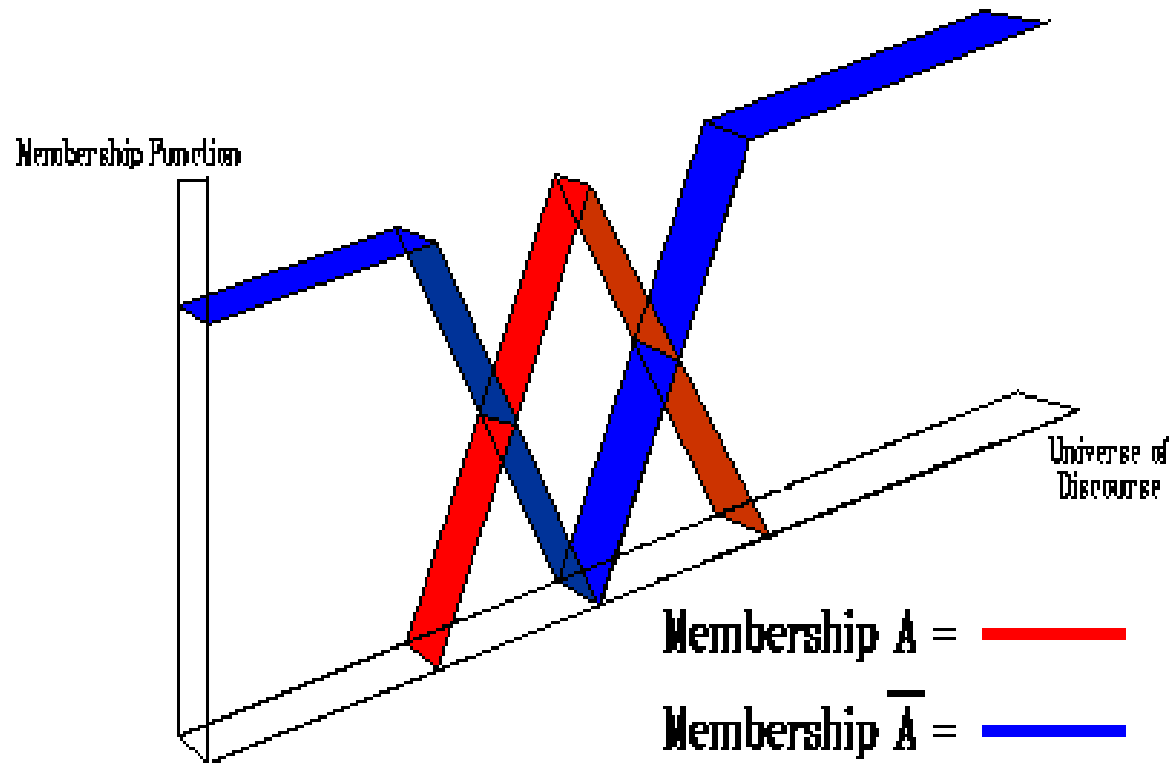


# Fuzzy Operators – OR



# Fuzzy Operators - Complement/NOT

$$\mu_{\bar{A}} = 1 - \mu_A$$



# Fuzzy Operators

- OR:
  - $A \text{ OR } B (x) = \text{MAX}(A(x), B(x))$  // Union.
  - $\sim A(x) = 1 - A(x)$ .
  - $A \text{ AND } B (x) = \text{MIN}(A(x), B(x))$  // Intersection.

# Algebraic Laws Still Apply

- De Morgans.
- Associativity:
  - $(A \text{ AND } B) \text{ AND } C = A \text{ AND } (B \text{ AND } C)$ .
- CommutativityL
  - $A \text{ AND } B = B \text{ AND } A$ .
- Distributivity:
  - $A \text{ AND } (B \text{ OR } C) = (A \text{ AND } B) \text{ OR } (A \text{ AND } C)$ .
- Etc...

# Hedges

- Hedges are linguistic modifiers on membership functions.
- Consider the membership function Is Tall(x).
- A hedge would be:
  - Is **Very** Tall(x).
  - Is **Sort Of** Tall(x).
  - Is **Quite** Tall(x).
  - Is **Slightly** Tall(x).
  - Is **Somewhat** Tall(x).

# Mathematical Representation of Hedges

- Consider a membership function  $\mu(x)$ :
  - A little =  $[\mu(x)]^{1.3}$
  - Slightly =  $[\mu(x)]^{1.7}$
  - Very =  $[\mu(x)]^2$
  - Extremely =  $[\mu(x)]^3$
  - More or Less, Somewhat =  $\text{SQRT}[\mu(x)]$

# Mathematical Representation of Hedges

- From the previous membership function for IsTall, consider  $\text{IsTall}(\text{John}) = \text{IsTall}(1.75)$ .
- This gives a fuzzy truth value of 0.8333.
- $\text{IsSomewhatTall} = \text{SQRT}(0.8333) = 0.913$ .
- $\text{IsVeryTall} = 0.8333^2 = 0.694$ .
- Etc...

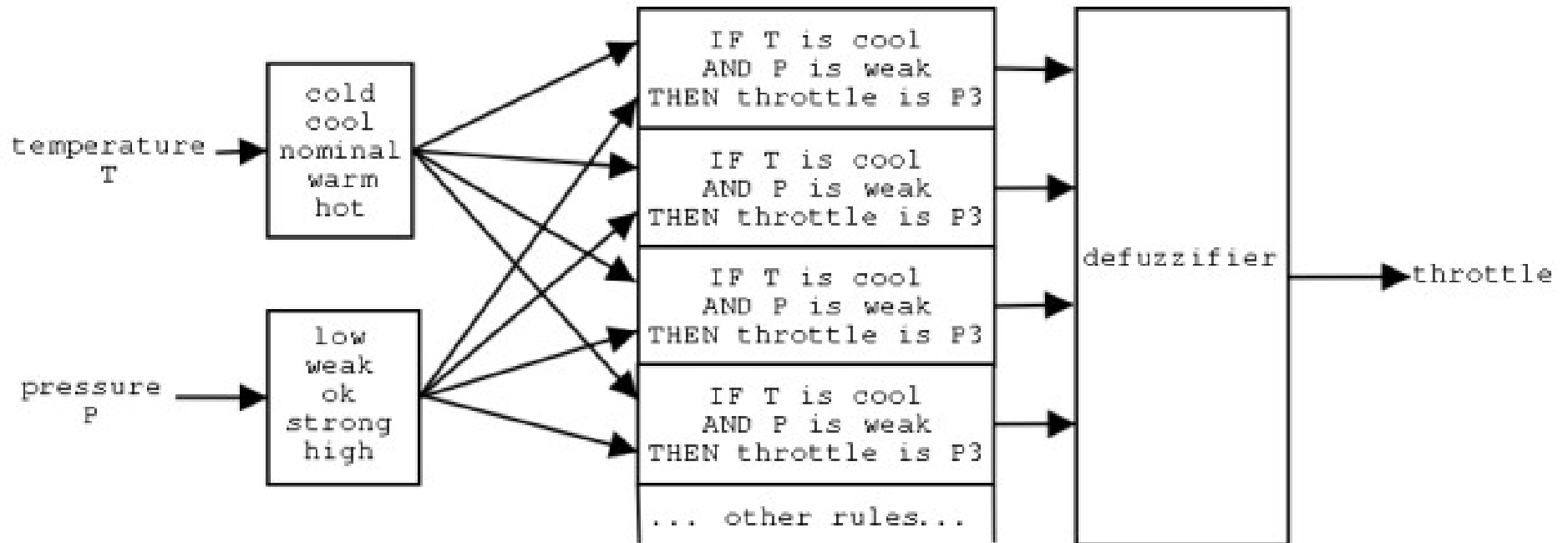
# Fuzzy Control

- Inputs from a sensor (e.g. Temperature, speed).
  - Possibility of ON/OFF (crisp 1, 0) inputs.
- Outputs to an actuator (e.g. Pressure).
- Wikipedia example:
  - IF BrakeTemperatureIsWarm AND SpeedIsNotVeryFast THEN BrakePressureIsSlightlyDecreased.
- There may be many fuzzy rules in a system (fuzzy rule set) and each rule may have conjunctives (AND, OR).

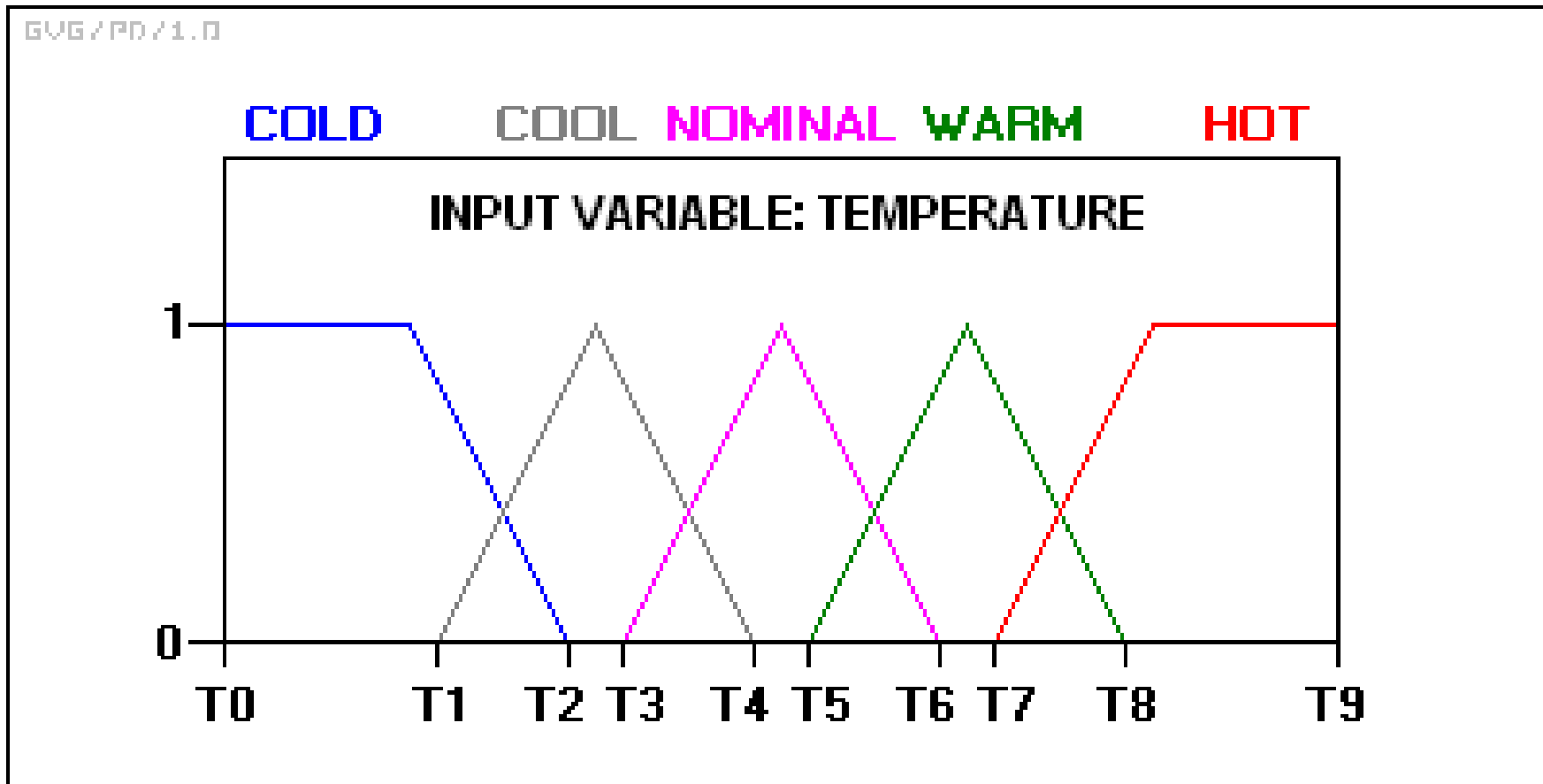


# Defuzzification

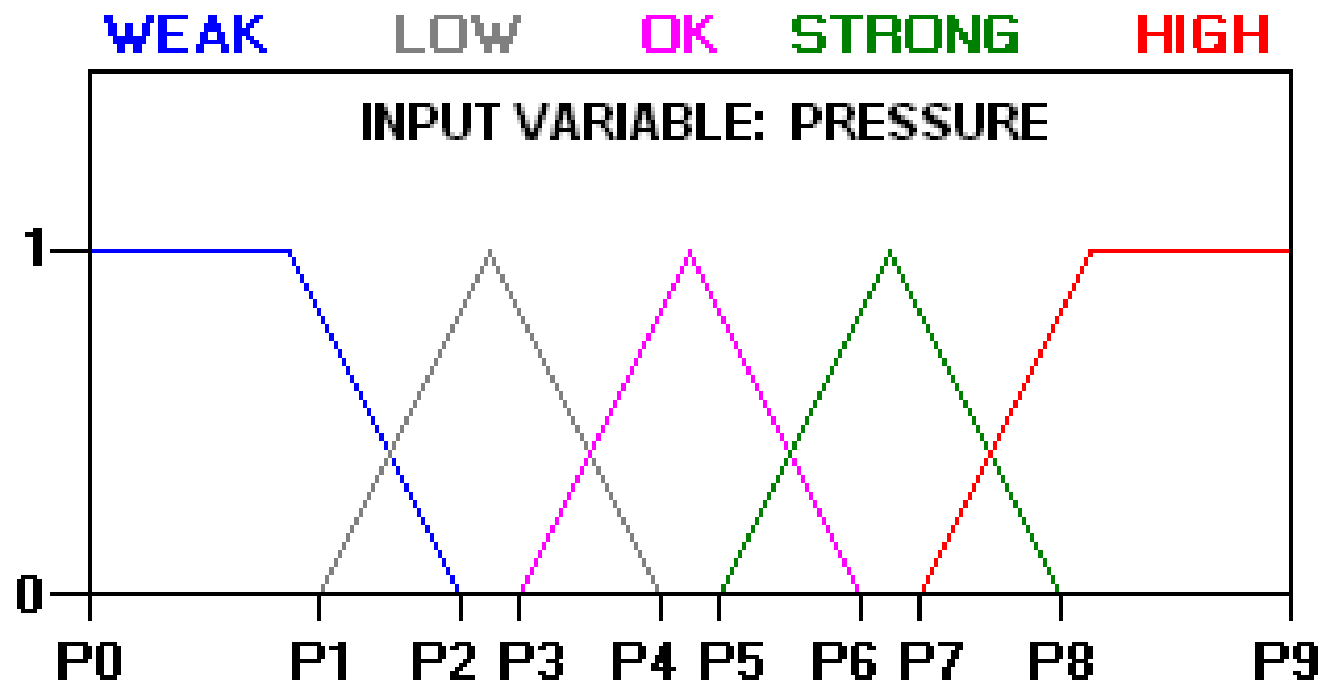
From [http://en.wikipedia.org/wiki/Fuzzy\\_control\\_system](http://en.wikipedia.org/wiki/Fuzzy_control_system)



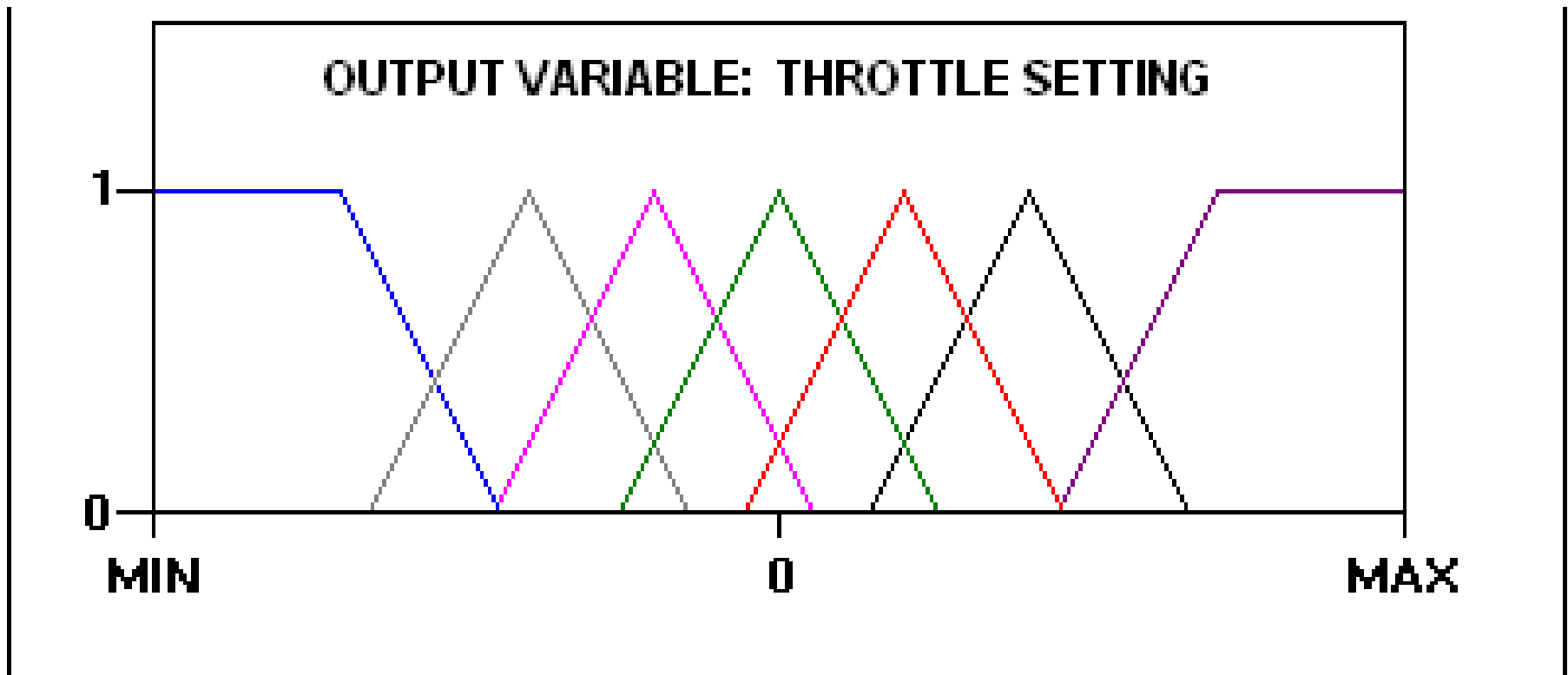
# Example (from Wikipedia)



# Example



# Example



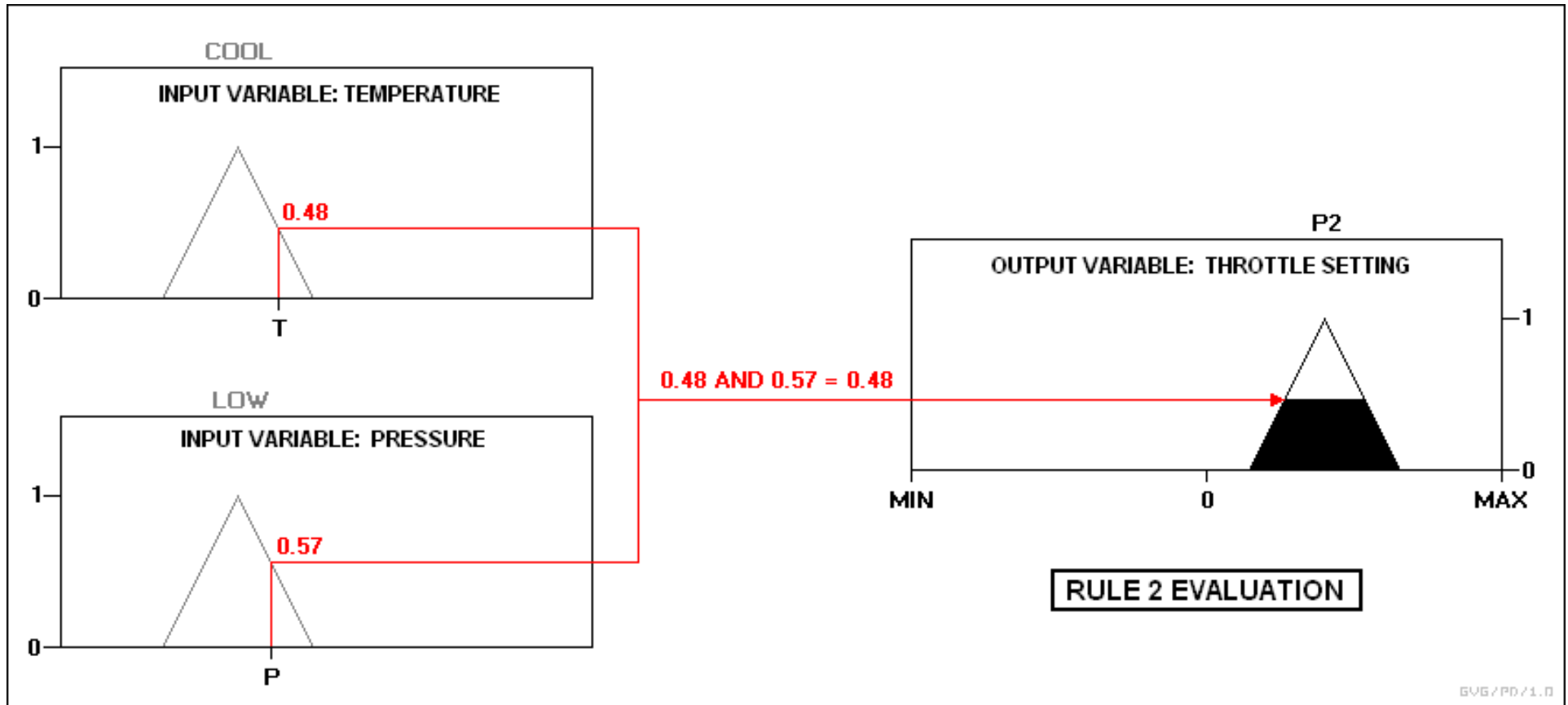
# Example Output Values

- N3 = Large Negative
- N2 = Medium Negative
- N1 = Small Negative
- Z = Zero
- P1 = Small Positive
- P2 = Medium Positive
- P3 = Large Positive

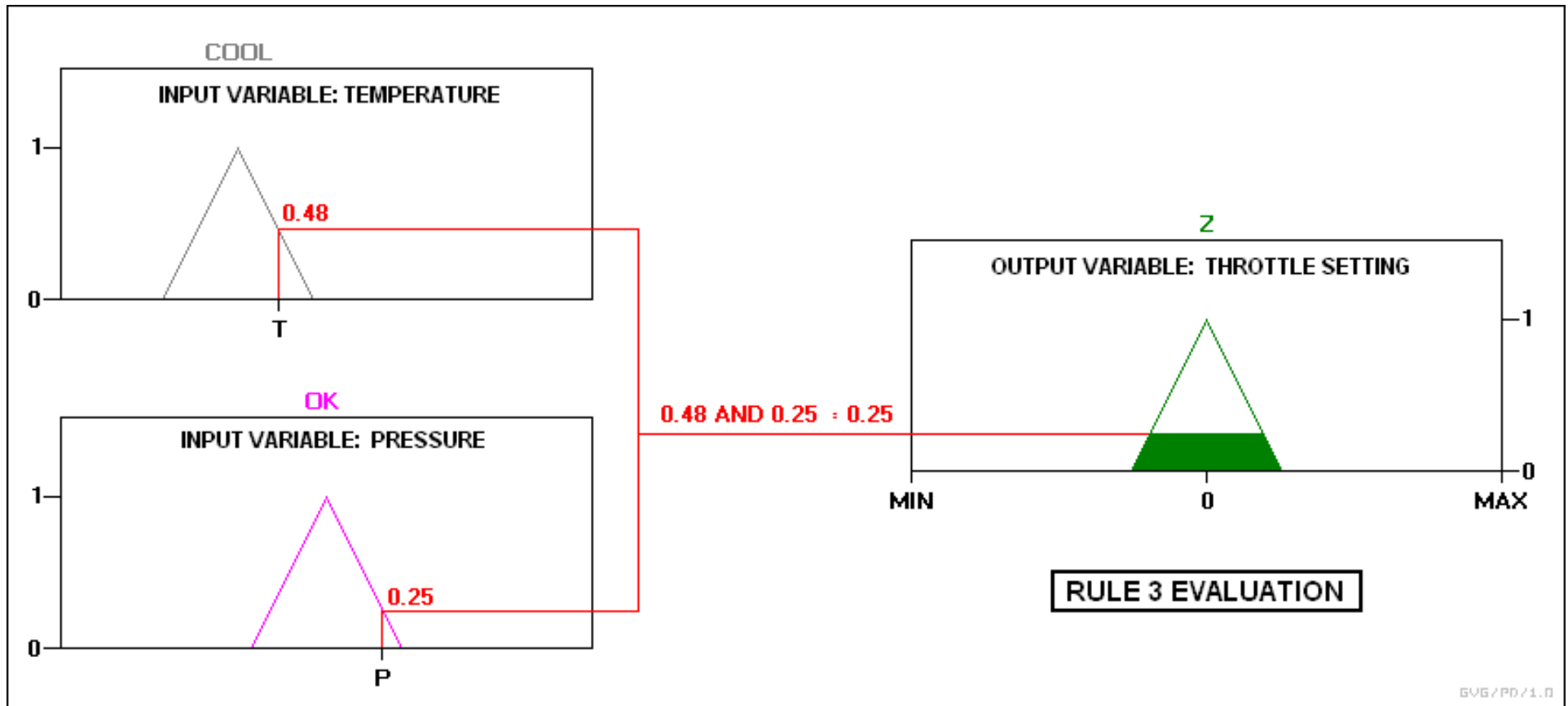
# Example Rules

- IF temperature IS cool AND pressure IS weak, THEN throttle is P3.
- IF temperature IS cool AND pressure IS low, THEN throttle is P2.
- IF temperature IS cool AND pressure IS ok, THEN throttle is Z.
- IF temperature IS cool AND pressure IS strong, THEN throttle is N2.
- Assume that the temperature is in the cool state and the pressure is in the low and ok states. Rules 2 and 3 will fire.

# Evaluation



# Evaluation

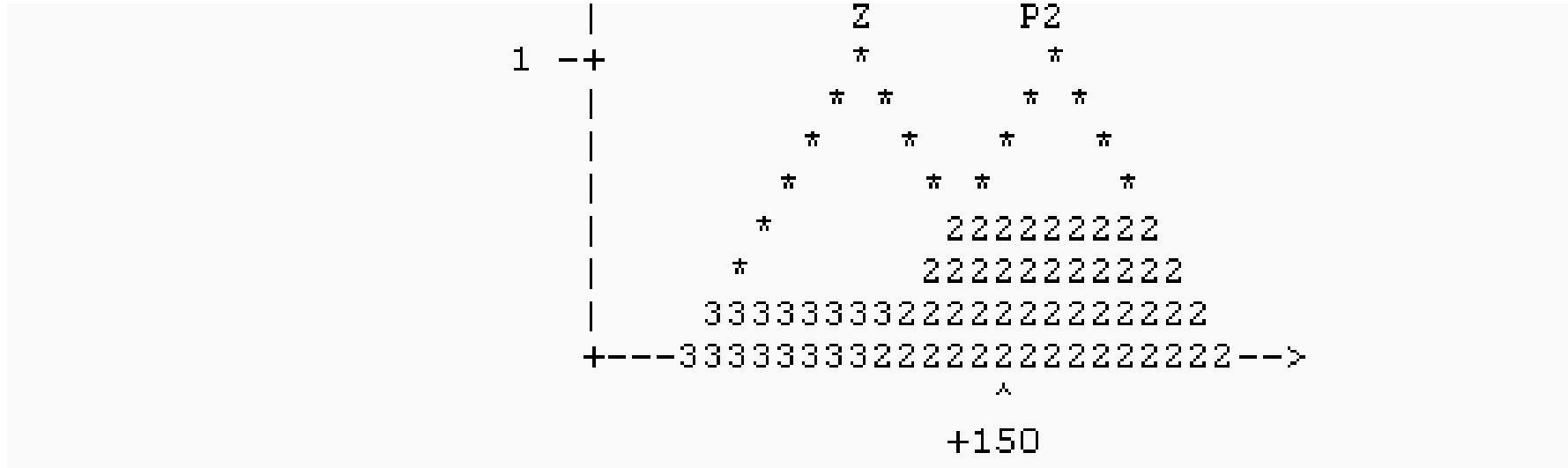




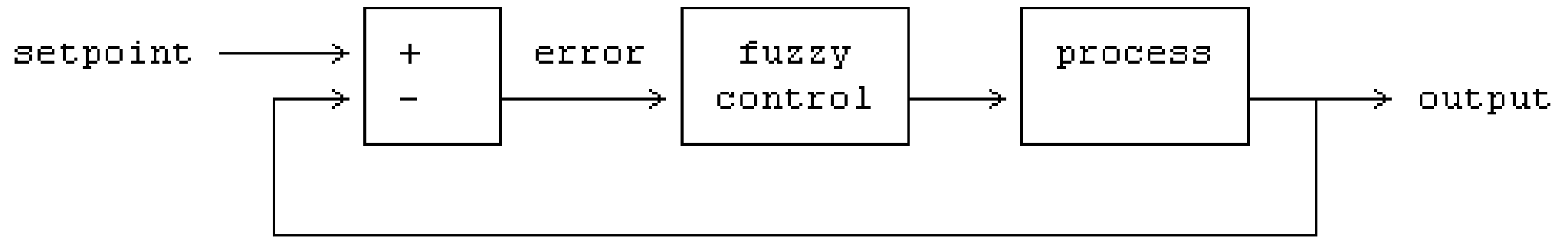
# Evaluation

- So, by rule 2, the pressure will be P2 (medium positive) with a truth value of 0.48.
- And by rule 3, the pressure will be Z (zero) with a truth value of 0.25.
- We now have to defuzzify the 2 rule outputs into one output value.
- Defuzzification can be done by either:
  - Choosing the maximum (P2 because it is 0.48, versus 0.25 for Z). This is not often used.
  - Choosing the centroid.

# Centroid Evaluation



# Controllers with Feedback



# Controllers with Feedback

- The system will generate:
  - Error (e).
  - Change in Error (delta).
- A rule base will exist that creates an output (repair) for the Error and Change In Error. E.g.:
  - If  $e = Z$  AND  $\text{delta} = Z$  THEN output = Z
  - If  $e = Z$  AND  $\text{delta} = SP$  THEN output = SN
  - If  $e = SN$  AND  $\text{delta} = SN$  THEN output = LP
  - If  $e = LP$  OR  $\text{delta} = LP$  THEN output = LN

# Error/Delta Table

	-1	-0.75	-0.5	-0.25	0	0.25	0.5	0.75	1
mu (LP)	0	0	0	0	0	0	0.3	0.7	1
mu (SP)	0	0	0	0	0.3	0.7	1	0.7	0.3
mu (ZE)	0	0	0.3	0.7	1	0.7	0.3	0	0
mu (SN)	0.3	0.7	1	0.7	0.3	0	0	0	0
mu (LN)	1	0.7	0.3	0	0	0	0	0	0

# Controllers with Feedback

- Suppose that at a point in time we have  $e = 0.25$ , and  $\delta = 0.5$ .
- By rule 1,  $\mu(1) = \text{MIN}(0.7, 0.3) = 0.3$ , output = 0.
- By rule 2,  $\mu(2) = \text{MIN}(0.7, 1) = 0.7$ , output = -0.25.
- By rule 3,  $\mu(3) = \text{MIN}(0, 0) = 0$ , output = 0.75.
- By rule 4,  $\mu(4) = \text{MAX}(0, 0.3) = 0.3$ , output = -1.

# Computing the Centroid

- For each rule  $x$ :
  - $\text{SUM}(\mu(x) * \text{output}(x))$

Divided by:

- $\text{SUM}(\mu(x))$

- In this case:

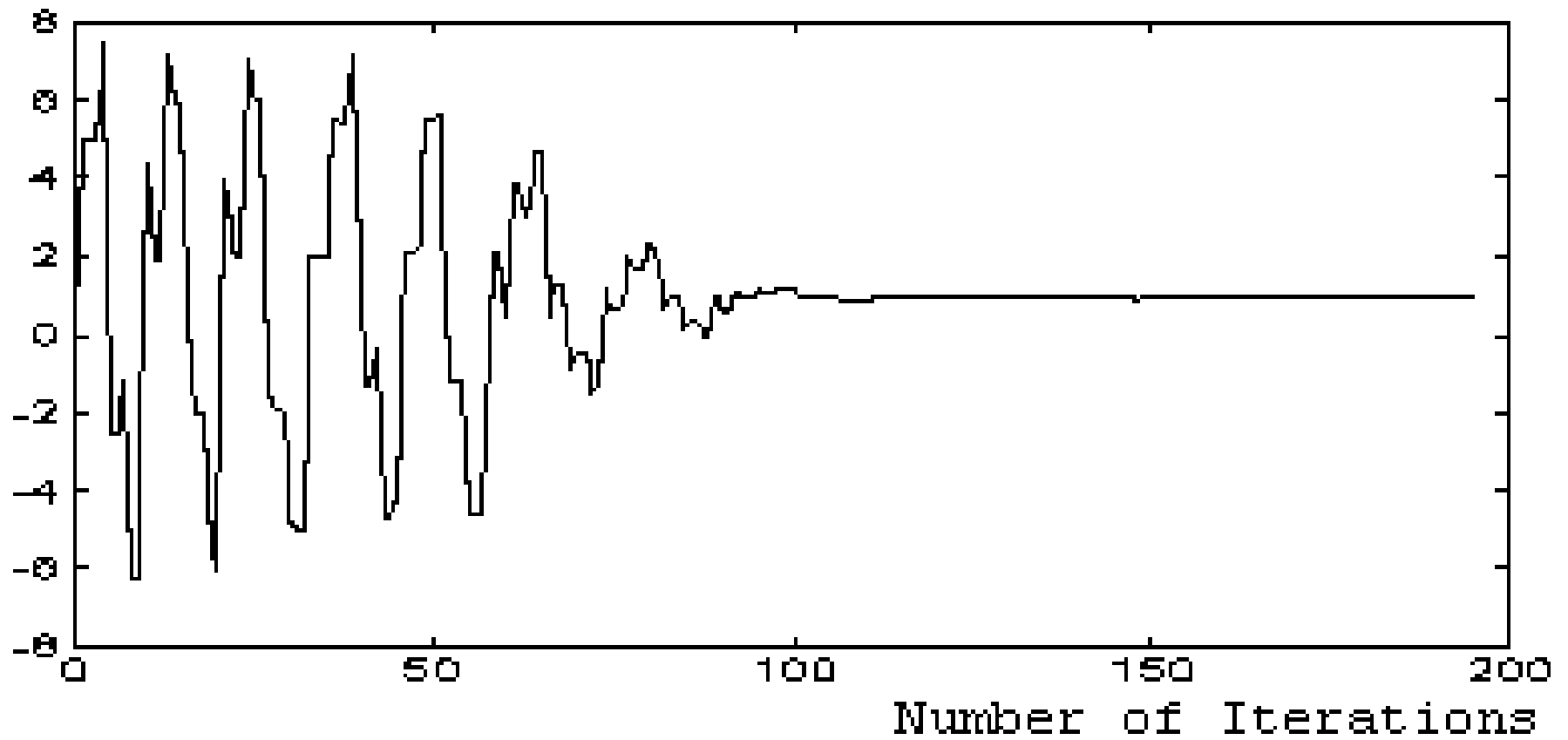
$$\frac{(0.3*0) + (0.7*-0.5) + (0*0.75) + (0.3*-1)}{0.3 + 0.7 + 0 + 0.3}$$

$$0.3 + 0.7 + 0 + 0.3$$

$$= -0.5$$

# Fluctuations

From [http://www.ici.ro/ici/revista/sic1997\\_2/](http://www.ici.ro/ici/revista/sic1997_2/)





# Advantages of Fuzzy Systems

- Do not require a mathematical model.
- Can fuzzify an expert system.
- Simple engine.