

Fuzzy logic

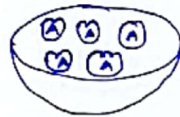


Dr. Mohammad Nurul Huda
M.Sc. in CSE Program
Neural Network and Fuzzy logic
Sheet-6

Fuzzy logic:

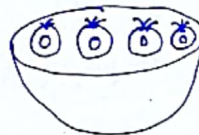
- A way to represent variation or imprecision in logic
- Fuzzy in the real-world sense "partly cloudy" - the distinctions that people use in decision-making all the time, but that computers and other advanced technology haven't been able to handle.

Crisp logic:



Is this a bowl of oranges?

Ans: NO

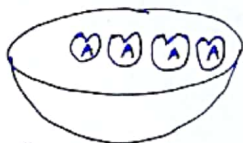


Is this a bowl of oranges?

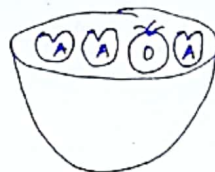
Ans: YES

Answer: {NO, YES}
: {0, 1}

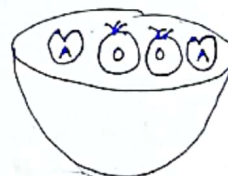
Thinking Fuzzy logic:



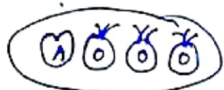
"Thinking fuzzy" about a bowl of oranges.



fuzzy bowl of apples



Fuzzy bowl of apples (continued)



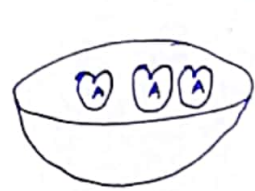
Fuzzy bowl of apples (continued)

Quest for Excellence Quest for Excellence Quest for Excellence Quest for Excellence Quest for Excellence
Quest for Excellence Quest for Excellence Quest for Excellence Quest for Excellence Quest for Excellence
Quest for Excellence Quest for Excellence Quest for Excellence Quest for Excellence Quest for Excellence

Characteristics of fuzziness:

1. Word based, not number based. For instance, hot; not 85°
2. Analog (ambiguous), not digital (YES/NO)
3. Nonlinear changeable.

Crisp values:



orange??

Ans: NO

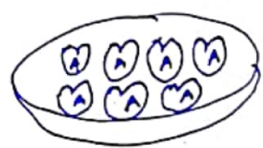


orange??

Ans: YES

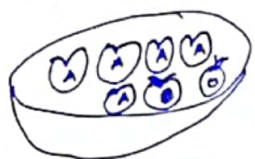
crisp set $\equiv \{NO, YES\} \equiv \{0, 1\}$

Fuzzy Values:



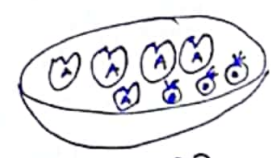
orange??

Ans: NO



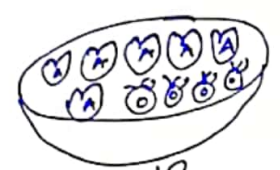
orange??

slightly



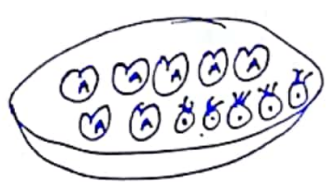
orange??

somewhat



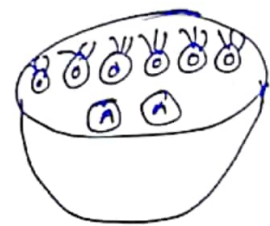
orange??

sort of

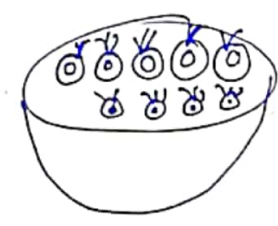


orange??

few



Mostly



YES, Absolutely.

Fuzzy set $\equiv \{no, slightly, somewhat, sort of, few, mostly, absolutely\}$

Fuzzy Words:

(3)

— Quantification:

all, most, many, ~~half~~ about half, few and no.

— Usuality:

Always, frequently, often, occasionally, seldom and never

— Likelihood:

certain, likely, uncertain, unlikely, and certainly not.

Fuzzy words are called linguistic variable.

Inventor of fuzzy logic:

Lotfi Zadeh in 1965.

Advantage of Fuzzy logic for System Control

- Fewer rules, values and decisions.
- Linguistic variables are used.
- Relation of input and output.
- Simplicity.
- Rapid prototyping.
- Easier to design.
- Increased Robustness.
- Simple knowledge representation.
- Few rules for great complexity.

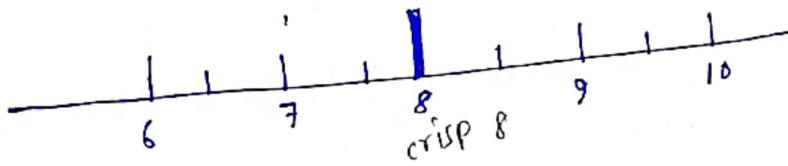
Drawbacks:

- Tuning of membership functions.
- May not scale well to large or complex problems.
- Deals with imprecision and vagueness, but not uncertainty.

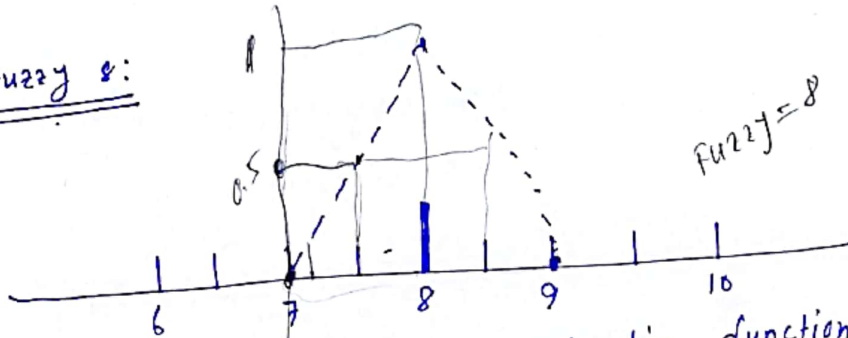
2y Numbers:

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A crisp 8:



A Fuzzy 8:



Eightness with a triangular Membership function:

Member	Degree of Membership
	0
7	0.5
7.5	1
8	0.5
8.5	0
9	

Traditional Representation of logic (Crisp logic):

```

boolean speed
get the speed
if speed = 0
    speed is slow
else
    speed is fast
    
```


Fuzzy Logic Representation:

float speed

get the speed

if $\text{speed} \geq 0.0$ and $\text{speed} < 0.25$
speed is slowest

else if $\text{speed} \geq 0.25$ and $\text{speed} < 0.5$
speed is slow

else if $\text{speed} \geq 0.5$ and $\text{speed} < 0.75$
speed is fast

else
speed is fastest

Crisp and Fuzzy arithmetic Operations:

Crisp

$$a = 3 \quad b = 2$$

Addition: $a + b$

$$3 + 2 = 5$$

subtraction: $a - b$

$$3 - 2 = 1$$

Multiplication: $a \times b$

$$3 \times 2 = 6$$

Division: ~~a/b~~ a/b

$$3/2 = 1.5$$

Fuzzy

$$a = (-2, \underline{3}, 8)$$

$$b = (-1, 2, 7)$$

$$\begin{array}{c} 10 \\ \downarrow \quad \downarrow \\ (-2, 3, 8) + (-1, 2, 7) \end{array}$$

$$= (5-9, 5, 5+9)$$

$$= (-4, 5, 14)$$

$$\frac{10+8}{2} = 9$$

$$\begin{array}{c} \downarrow \quad \downarrow \\ (-2, 3, 8) - (-1, 2, 7) \end{array}$$

$$= (1-9, 1, 1+9)$$

$$= (-8, 1, 10)$$

$$\begin{array}{c} \downarrow \quad \downarrow \\ (-2, 3, 8) \times (-1, 2, 7) \end{array}$$

$$= (6-9, 6, 6+9)$$

$$= (-3, 6, 15)$$

$$\begin{array}{c} \downarrow \quad \downarrow \\ (-2, 3, 8) / (-1, 2, 7) \end{array}$$

$$= (1.5-9, 1.5, 1.5+9)$$

$$= (-7.5, 1.5, 10.5)$$

different types of membership function:

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Triangular:

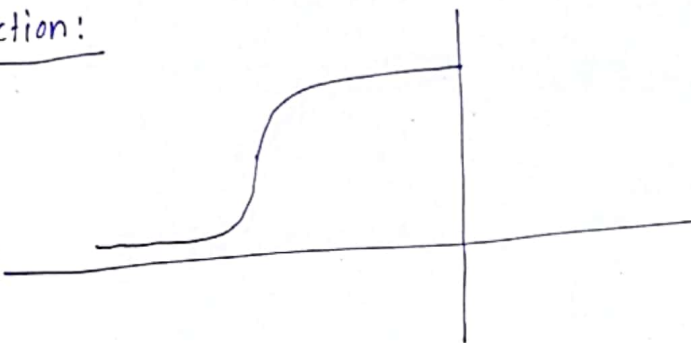


Trapezoid:

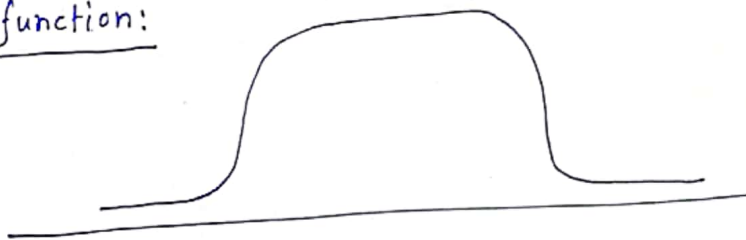


S-function:

S function

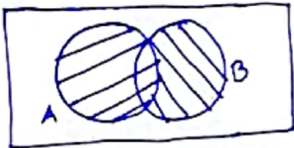
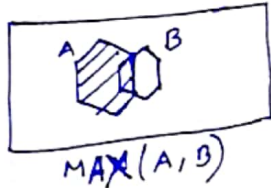
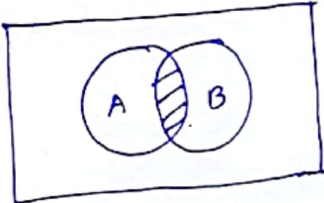
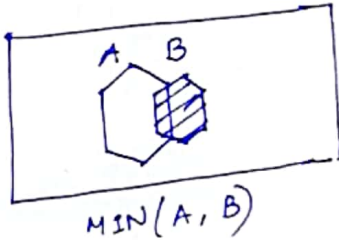
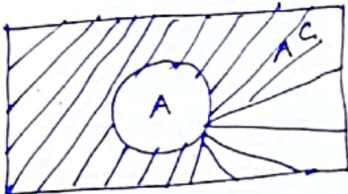
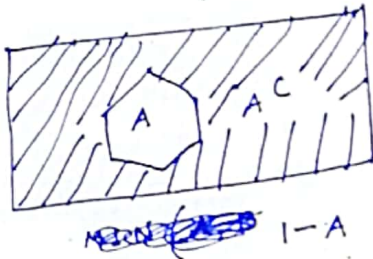
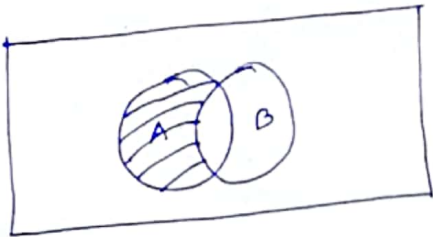
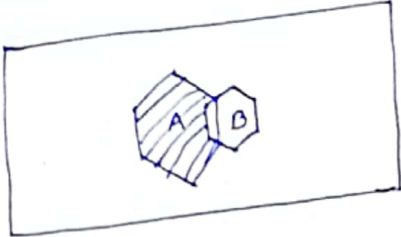



pi-function:



z-function:



Operation	crisp	Fuzzy
Disjunction or \vee or \cup		 $\text{MAX}(A, B)$
Conjunction or intersection		 $\text{MIN}(A, B)$
Complement		 $\text{MIN}(A, B)$ $1 - A$
Difference	 $A - B$	 $A - B$ $A \rightarrow B$
Implication	 A implies B or, $A \rightarrow B$	

Fuzzy logic calculator:

$$0.3 \wedge 0.8 = \text{MIN}(0.3, 0.8) = 0.3$$

$$0.2 \vee 0.6 = \text{MAX}(0.2, 0.6) = 0.6$$

$$\neg 0.9 = 1.0 - 0.9 = 0.1$$

$$\neg 0.6 = 1.0 - 0.6 = 0.4$$

$$0.8 \Rightarrow 0.7 \quad 0.8 > 0.7$$

$$0.7 \Rightarrow 0.8 \quad 0.7 > 0.8$$

YES
NO

Fuzzy logic in Multielement sets:

$$A \equiv (0.8 \quad 0.2 \quad 0.7)$$

$$B \equiv (1 \quad 0.3 \quad 0.4)$$

Operation	Fuzzy logic
$A \cup B$	$(1.0 \quad 0.3 \quad 0.7)$
$A \cap B$	$(0.8 \quad 0.2 \quad 0.4)$
$B \setminus A$	$(\text{NO} \quad \text{NO} \quad \text{YES})$
$A \setminus B$	$(0 \quad 0 \quad 3)$
A^c	$(0.2 \quad 0.8 \quad 0.3)$

Crisp set:

$$A \equiv (0.8 \quad 0.2 \quad 0.7)$$

$$B \equiv (1 \quad 0.3 \quad 0.4)$$

$$A \equiv (1 \quad 0 \quad 1)$$

$$B \equiv (1 \quad 0 \quad 0)$$

fuzzy set

crisp set

and Fuzzy logic (Rules of inference)

Set theory is closely related to the truth-finding logical statements called the rules of inference.

⇒ Use rules of implication ($A \Rightarrow B$)

→ Modus ponens: (Affirmative mode)

If the apple is red
AND a red apple is a ripe apple
Then the apple is ripe

→ Modus tollens: (Denial mode)

If the apple is not ripe
AND a red apple is a ripe apple
Then the apple is not red

Fuzzy Rule 1: (Modus ponens)

As the apple is very red
And a red apple is a ripe apple
Then the apple is very ripe

Fuzzy Rule 2: (Compositional rule of inference)

As Apple #1 is very ripe
And Apple #2 is not quite as ripe as Apple #1
Then Apple #2 is more or less ripe.

logical statements

Existential quantifier

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isp:

$$(\exists x) [\text{ripe}(\text{apple})]$$

There exists one example of a ripe apple.

Fuzzy:

- There exists
- Few

crisp:

$$(\forall x) [\text{apple}(x) \rightarrow \text{ripeness}(x)]$$

All apples are ripe

Fuzzy

- For all
- Most

Traditional or Crisp Rule :

Ex: If the room temperature is less than 62 degrees
Then set the thermostat for 68 degrees

* Crisp rule gives precise terms.

Fuzzy logic

If-Then-style rules expressed by As-Then (general form) or As-Do (control form).

Ex: As the room temperature is cool
Do turn on the heater to High

Fuzzy logic NOT

A	Not(A)
0	1
0.25	0.75
0.5	0.5
0.75	0.25
1	0

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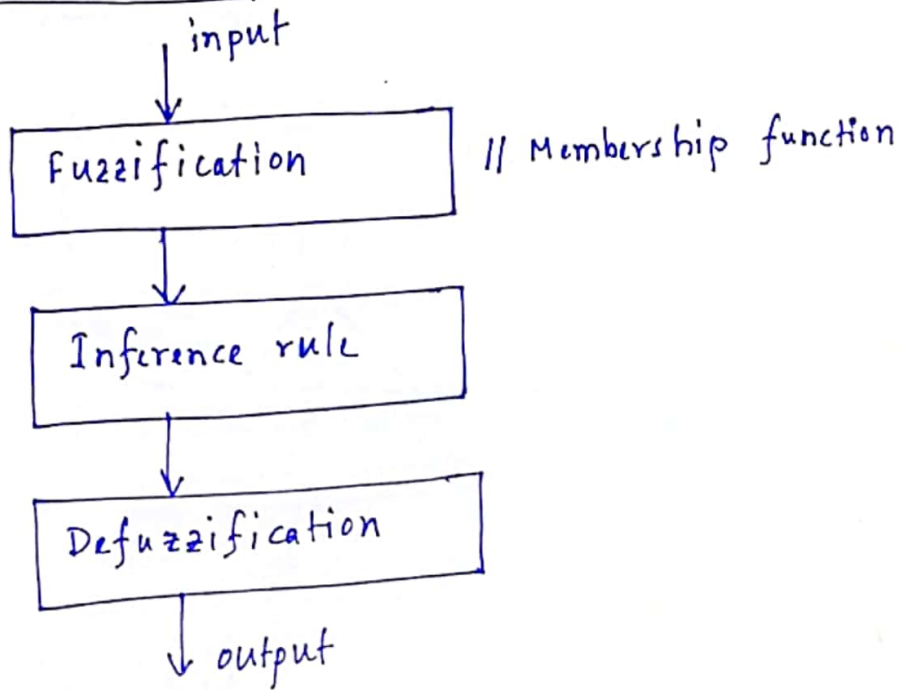
Fuzzy logic AND

A \ B	0	0.25	0.5	0.75	1.0
0	0	0	0	0	0
0.25	0	0.25	0.25	0.25	0.25
0.5	0	0.25	0.5	0.5	0.5
0.75	0	0.25	0.5	0.75	0.75
1.0	0	0.25	0.5	0.75	1.0

Fuzzy logic OR

A \ B	0	0.25	0.5	0.75	1.0
0	0	0.25	0.5	0.75	1.0
0.25	0.25	0.25	0.5	0.75	1.0
0.5	0.5	0.5	0.5	0.75	1.0
0.75	0.75	0.75	0.75	0.75	1.0
1.0	1.0	1.0	1.0	1.0	1.0

Logic Systems:



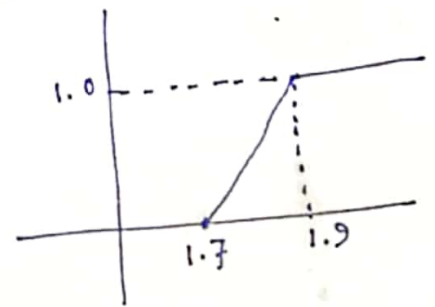
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Membership Functions:

Better than listing membership values.

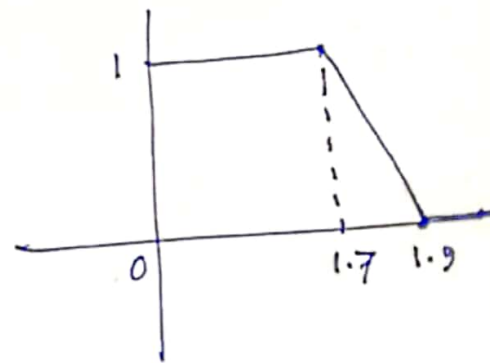
Example: Fuzzy tall

$$\text{Tall}(x) = \begin{cases} 1 & \text{if } x \geq 1.9 \text{ m} \\ 0 & \text{if } x \leq 1.7 \text{ m} \\ \frac{x-1.7}{0.2} & \text{otherwise} \end{cases}$$



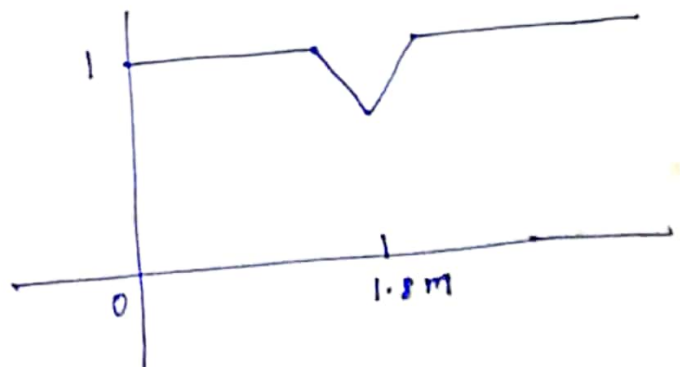
Example: Fuzzy short

$$\text{short}(x) = \begin{cases} 0 & \text{if } x \geq 1.9 \text{ m} \\ 1 & \text{if } x \leq 1.7 \text{ m} \\ \frac{1.9-x}{0.2} & \text{otherwise} \end{cases}$$



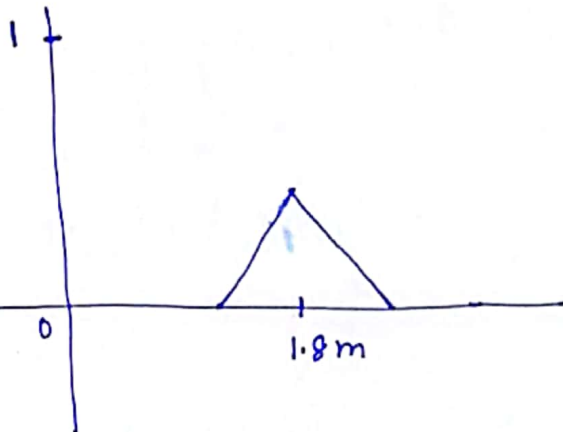
Union:

$\text{Tall}(x) \cup \text{short}(x)$:



function:

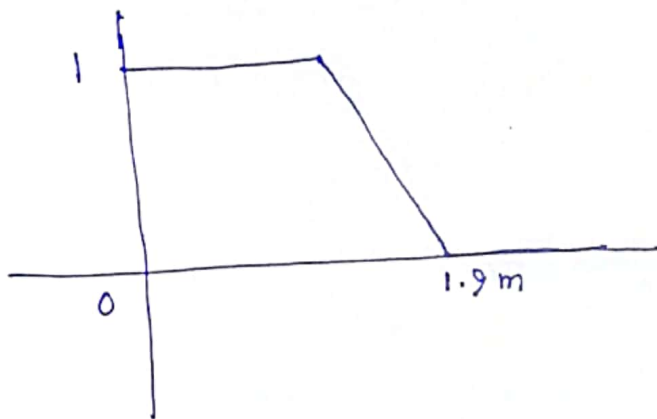
Tall(x) and short(x)



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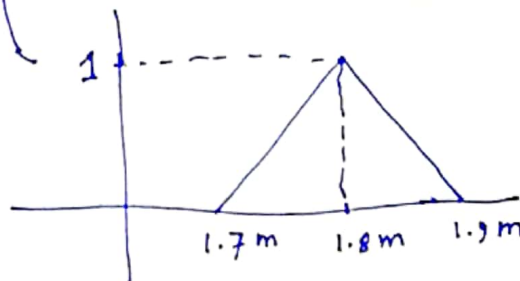
complement:

Not(Tall(x))



Example:

$$\text{Medium}(x) = \begin{cases} 0 & \text{if } x \geq 1.90 \text{ or } x < 1.70 \\ \frac{1.9 - x}{0.1} & \text{if } x \geq 1.80 \text{ and } x < 1.90 \\ \frac{x - 1.70}{0.1} & \text{if } x \geq 1.70 \text{ and } x < 1.80 \end{cases}$$



Input: Temperature { Freezing, Cool, Warm, Hot }
 cloud Cover { Sunny, ~~Cloud~~, Partly, Overcast }

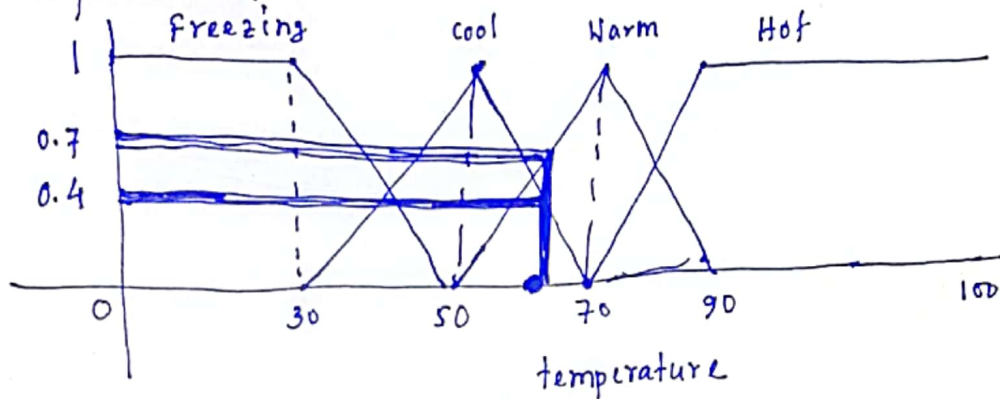
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Output: Speed { slow, Fast }

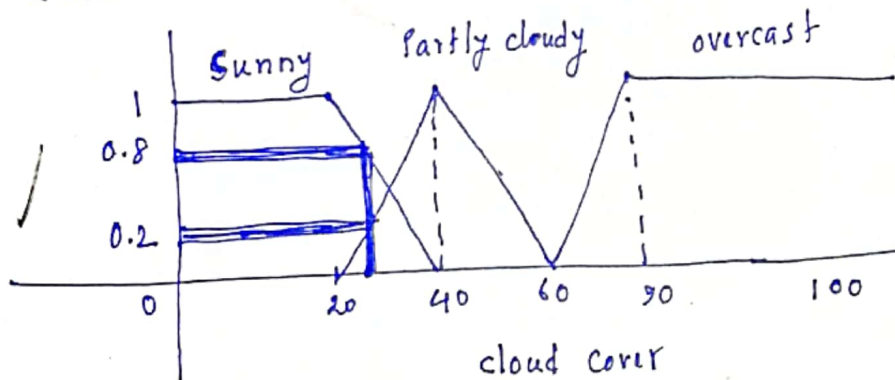
fast will I go if it is 65°F and 25% cloud cover?

membership function / Fuzzification:

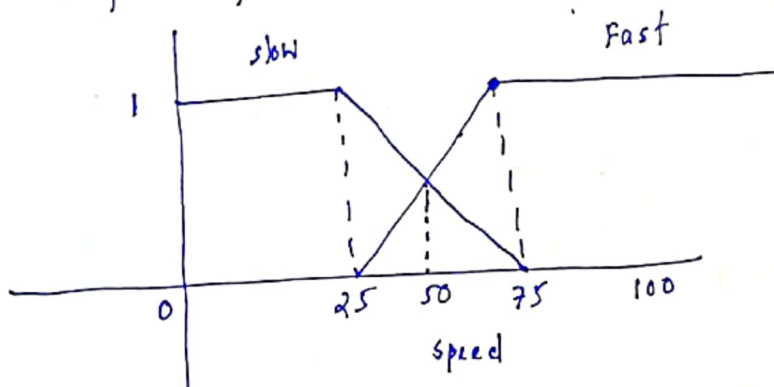
For Temperature:



For cloud Cover:



For output speed:



$65^{\circ}\text{F} \Rightarrow \text{cool} = 0.4, \text{Warm} = 0.7$
 $25\% \text{ cover} \Rightarrow \text{Sunny} = 0.8, \text{cloudy} = 0.2$

is sunny and warm, drive fast

it is cloudy and cool, drive slow

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sunny and warm = ~~0.8 and 0.7~~ sunny \wedge warm

$$= 0.8 \wedge 0.7$$

$$= 0.7$$

By rule 1, Fast = $0.7 = 70\%$

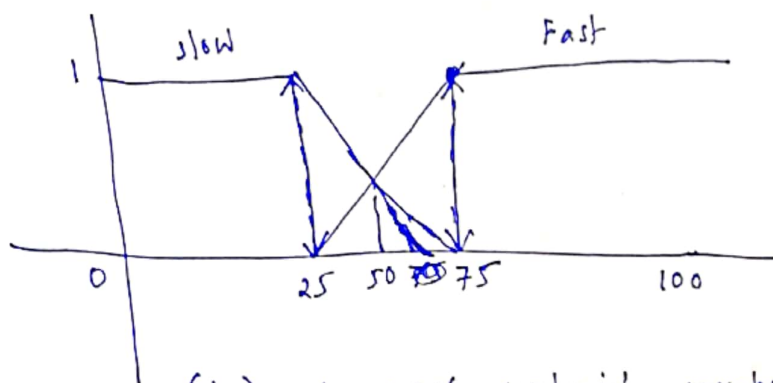
cloudy and cool = cloudy \wedge cool

$$= 0.2 \wedge 0.4$$

$$= 0.2$$

By rule 2, slow = $0.2 = 20\%$

Defuzzification: (output speed)



(slow) At 25% centroid, membership is 100%.

(Fast) At 75% centroid, membership is 100%.

Generated speed = Weighted Mean

$$= (2 \times 25 + 7 \times 75) / (2 + 7) = 63.8 \text{ mph } \underline{\underline{\text{Ans!}}}$$