

# Fuzzy Propositions



# Fuzzy Propositions

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## OBJECTIVES

1. Introduce fuzzy propositions.
2. Learn to evaluate truth value of a fuzzy proposition.
3. Understand fuzzy quantifiers
4. Understand linguistic hedges



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# Fuzzy Propositions

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➤ The truth value of a fuzzy proposition is a matter of degree expressed by a fraction in the unit interval  $[0, 1]$ . For example,

“Tina is young”

is a fuzzy proposition. Depending upon the age of Tina the truth value of the proposition assume any value, say 0.63.

➤ Simple fuzzy propositions may be classified into four types:

1. unconditional and unqualified proposition
2. unconditional and qualified proposition
3. conditional and unqualified proposition
4. conditional and qualified proposition



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**1. Unconditional and unqualified proposition:** The canonical form of this type of proposition,  $p$  is shown below:

$p: \mathcal{V} \text{ is } F.$

where,  $\mathcal{V}$  is a variable on which fuzzy set  $F$  is defined.

Example of such a proposition is:

$p: \text{temperature is high.}$

Here, *temperature* is the variable  $\mathcal{V}$  and *high* is the fuzzy set  $F$ .

Truth value,  $T(p)$  of such a proposition is given by membership grade of the particular value,  $v$  of temperature in the fuzzy set  $F$ . That is,

$$T(p) = F(v).$$

This is illustrated by the example in Fig. 1.

# Fuzzy Propositions

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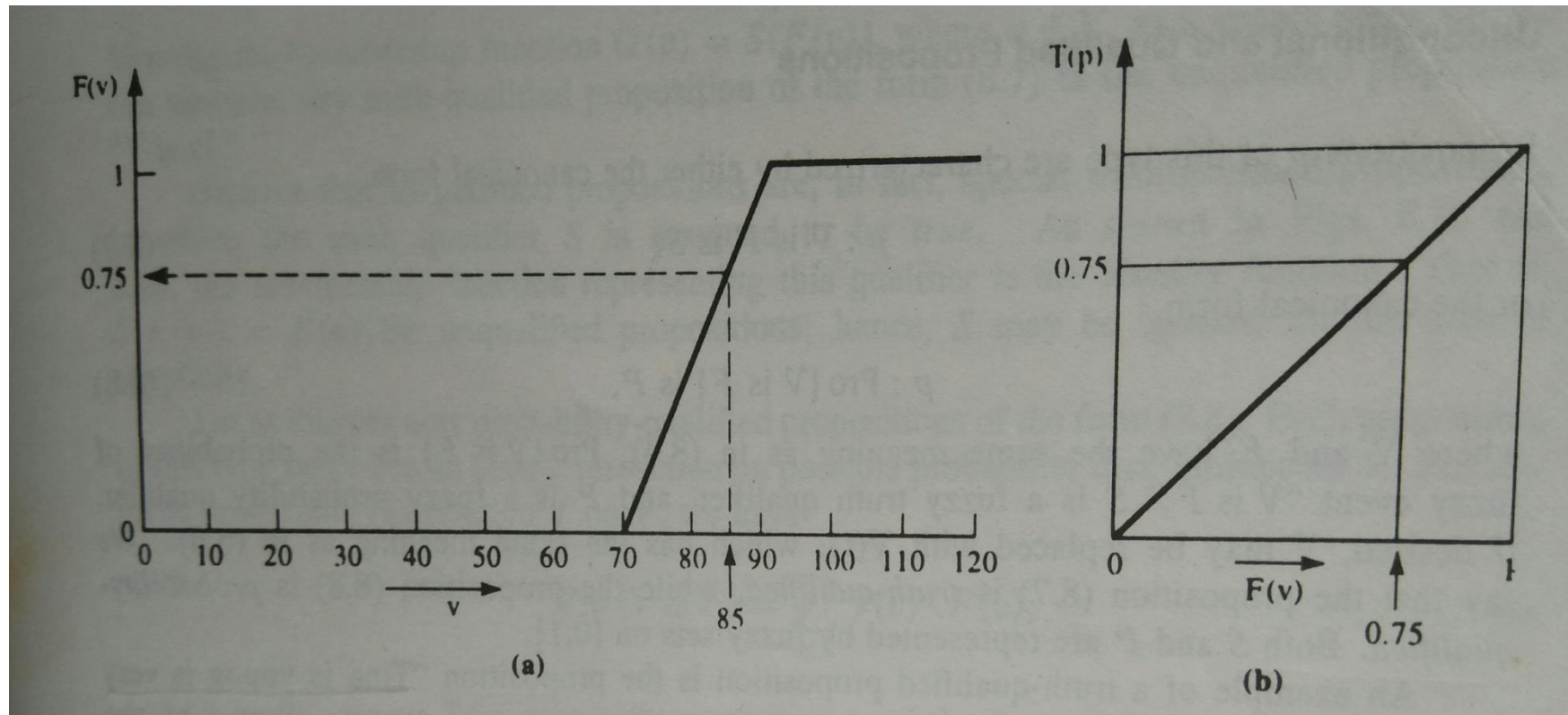


Fig. 1. Components of fuzzy proposition:  $p$ : Temperature( $v$ ) is high ( $F$ ).

**2. Unconditional and qualified proposition:** Canonical form of this type of proposition is,

$p: V \text{ is } F \text{ is } S.$

Where  $S$  is a fuzzy truth qualifier.

An example of such a truth-qualified proposition is:

$p: \text{Tina is young is very true.}$

Here, very is the truth qualifier represented by  $S$ .

The degree of truth,  $T(p)$ , of any truth-qualified proposition  $p$  is given for each value  $v$  of the underlying variable by the equation

$$T(p) = S(F(v))$$

The process can be understood with Fig. 2.

# Fuzzy Propositions

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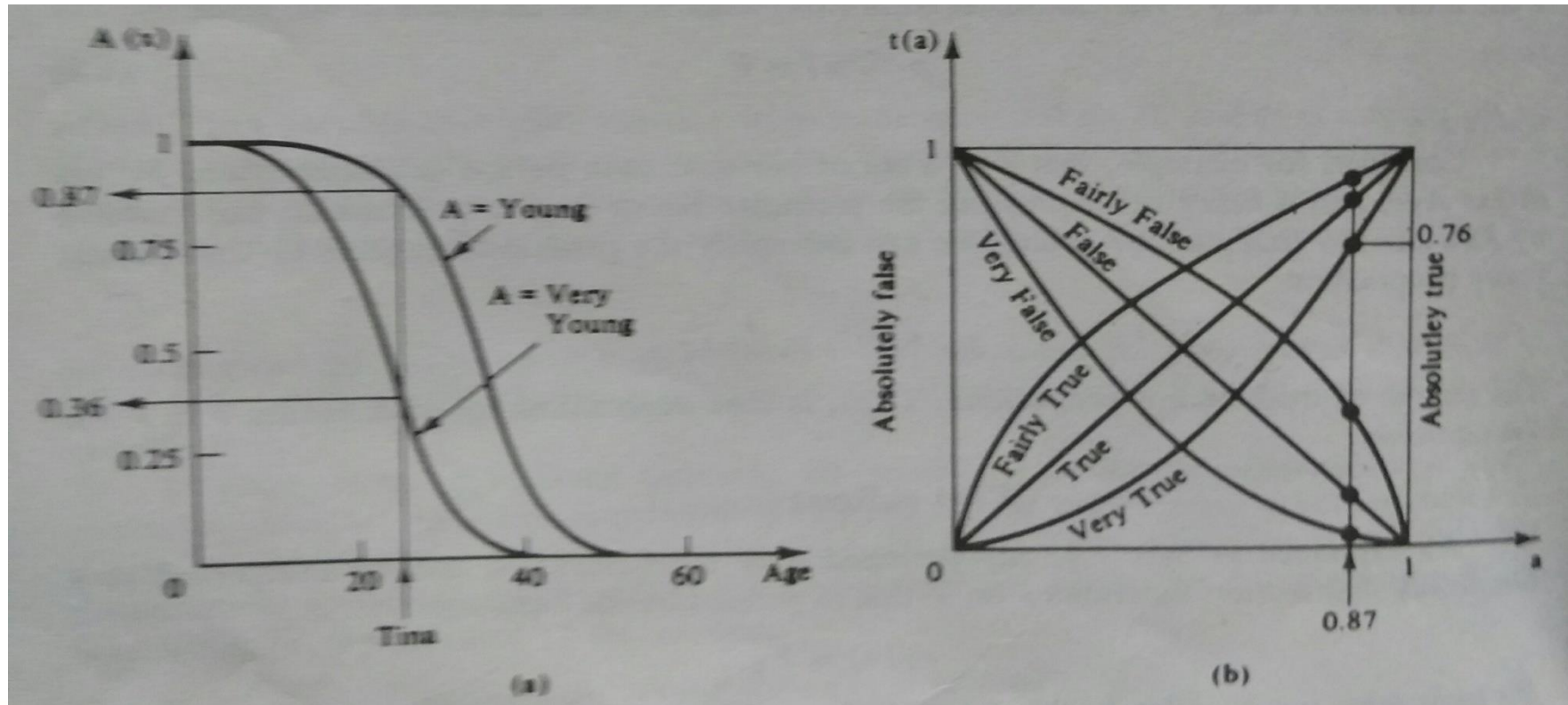


Fig. 2 Truth values of a fuzzy proposition



**3. Conditional and unqualified proposition:** Proposition  $p$  of this type are expressed by the canonical form

$p$ : If  $\mathcal{X}$  is  $A$ , then  $\mathcal{Y}$  is  $B$ ,

whrer  $\mathcal{X}$  and  $\mathcal{Y}$  are variables whose values are in sets  $X$  and  $Y$ , respectively, and  $A$  and  $B$  are fuzzy sets on  $X$  and  $Y$ . These propositions may also be viewed as propositions of the form

$(\mathcal{X}, \mathcal{Y})$  is  $R$ ,

where  $R$  is a fuzzy relation that is determined for each  $x \in X$  and  $y \in Y$  by the formula

$$R(x, y) = \partial(A(x), B(y))$$

here,  $\partial$  denotes a suitable fuzzy implication like Lukasiewicz implication

$$\partial(a, b) = \min(1, 1 - a + b)$$



**4. Conditional and qualified proposition:** One canonical form of this type of proposition is

$p$ : If  $X$  is  $A$ , then  $Y$  is  $B$  is  $S$

Truth value of such a proposition can be evaluated by combining the methods used for the previous types of propositions.

**Fuzzy quantifiers:** Fuzzy quantifiers are fuzzy numbers that take part in fuzzy propositions. There are two kinds.

1. **First kind:** Fuzzy quantifiers of this kind are defined on the set of real numbers and characterize linguistic terms such as *about 10*, *much more than 100*, *at least about 5*, and so on.
2. **Second kind:** Fuzzy quantifiers of this kind are defined on  $[0, 1]$  and characterize linguistic terms such as *almost all*, *about half*, *most* etc.

Examples of fuzzy quantifiers of the second kind are shown in Fig. 3.

# Fuzzy Quantifiers

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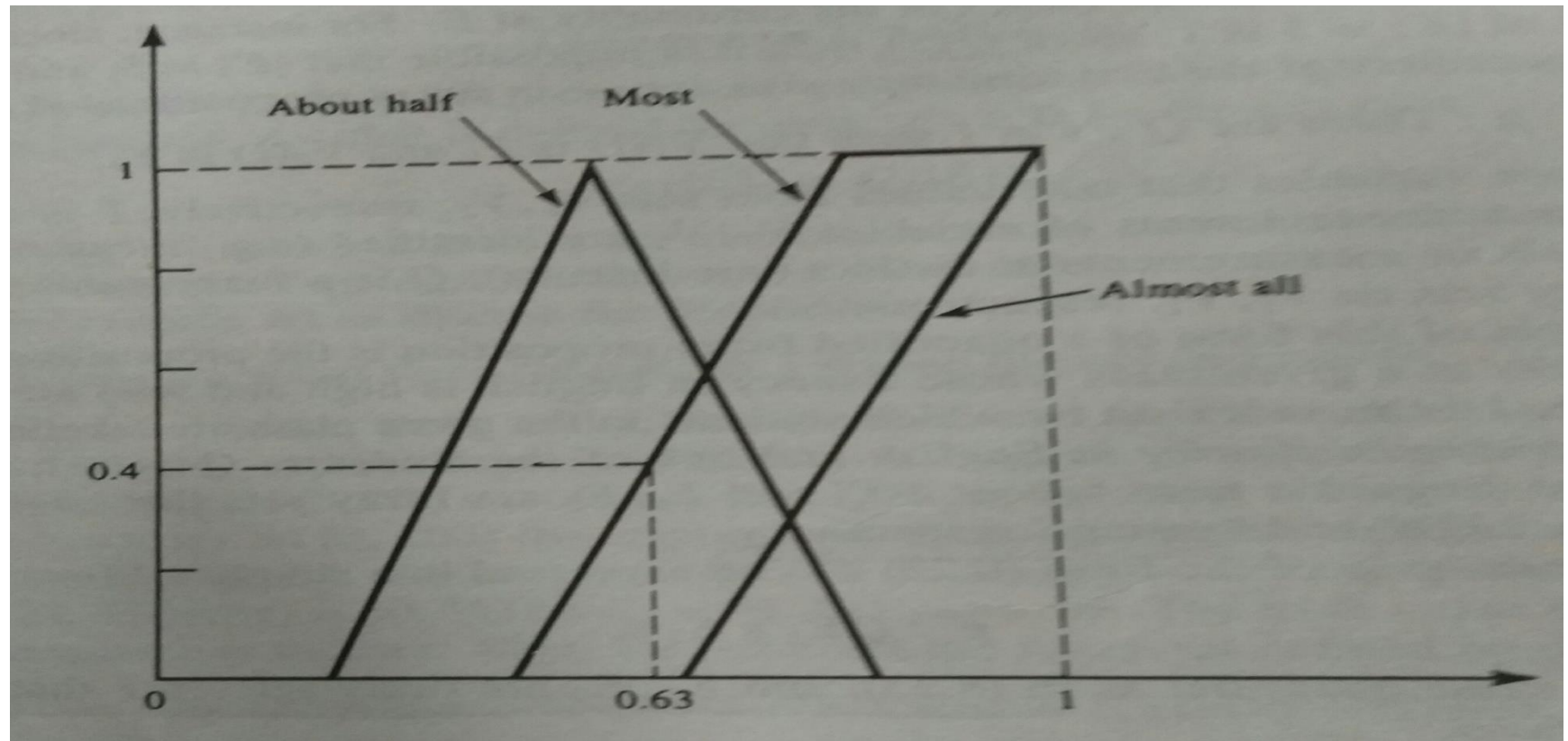


Fig. 3 Fuzzy quantifiers of the second kind.



- **Linguistic Hedges:** Linguistic hedges ( or simply hedges) are special linguistic terms by which other linguistic terms are modified.
- Example: Linguistic terms such as *very*, *more or less*, *fairly*, *extremely* etc.
- Hedges can be used to modify fuzzy predicates, fuzzy truth values, and fuzzy probabilities. For example,
- - “x is very young is true”
  - “x is young is very true”
  - “x is very young is very true”

- Any linguistic hedge,  $H$ , may be interpreted as a unary operation,  $h$ , called a modifier, on the interval  $[0, 1]$ .
- For example, the hedge *very* is often interpreted as  $h(a)=a^2$ , the hedge *fairly* is interpreted as  $h(a)=\sqrt{a}$  ( $a \in [0, 1]$ ).
- The modifier  $h$  is called strong if  $h(a) < a$  for all  $a \in [0, 1]$ , the modifier called weak if  $h(a) > a$  for all  $a \in [0, 1]$ . The special modifier  $h(a)=a$  is called an identity modifier.
- A strong modifier strengthens a fuzzy predicate to which it is applied, and consequently it reduces the truth value of the associated proposition. Opposite is the case with weak modifiers.

For example:

$p_1$ : John is young,

$p_2$ : John is very young,

$p_3$ : John is fairly young.

Let, the linguistic hedges *very* and *fairly* be represented by strong modifier  $\alpha^2$  and weak modifier  $\sqrt{\alpha}$ . Assume that John is 26, and  $Y$  represents the fuzzy set *young*,  $Y(26)=0.8$ . Then,

$\text{very young}(26) = 0.8^2 = 0.64,$

$\text{fairly young}(26) = \sqrt{0.8} = 0.89.$

A convenient class of function that can be used as modifier is

$h_\alpha(a) = a^\alpha$ , where  $\alpha$  is a parameter whose value is a +ve real number.

When  $\alpha < 1$ ,  $h_\alpha$  is a weak modifier, when  $\alpha > 1$ ,  $h_\alpha$  is a strong modifier.

Another example:

Suppose we define fuzzy sets *Small* and *Large* on  $Y=\{1,2,3,4,5\}$  as:

$$\text{Small} = \frac{1}{1} + \frac{0.8}{2} + \frac{0.6}{3} + \frac{0.4}{4} + \frac{0.2}{5} \quad \text{Large} = \frac{0.2}{1} + \frac{0.4}{2} + \frac{0.6}{3} + \frac{0.8}{4} + \frac{1}{5}$$

Obtain the fuzzy sets representing:

- i) *Very small*
- ii) *Very very large*

Solution:

$$\begin{aligned} \text{very small} &= \frac{1^2}{1} + \frac{0.8^2}{2} + \frac{0.6^2}{3} + \frac{0.4^2}{4} + \frac{0.2^2}{5} \\ &= \frac{1}{1} + \frac{0.64}{2} + \frac{0.36}{3} + \frac{0.16}{4} + \frac{0.04}{5} \end{aligned}$$

$$\begin{aligned} \text{very very large} &= \frac{(.2^2)^2}{1} + \frac{(.4^2)^2}{2} + \frac{(.6^2)^2}{3} + \frac{(.8^2)^2}{4} + \frac{(1^2)^2}{5} \\ &= \frac{.0016}{1} + \frac{.0256}{2} + \frac{.1296}{3} + \frac{.4096}{4} + \frac{1}{5} \end{aligned}$$



# Fuzzy Inference



# Introduction

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- Fuzzy inference is the process of obtaining new knowledge through existing knowledge.
- Knowledge is most commonly represented in the form of rules or proposition for example “if x is A then y is B” (Where A and B are linguistic values defined by fuzzy sets on universes of discourse X and Y). A rule is also called a fuzzy implication.
- “x is A” is called the antecedent or premise and “y is B” is called the consequence or conclusion.



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# Important Inferring Processes

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- Generalized modus Ponens (GMP) - Latin for “The way that affirms by affirming”
- Generalized modus Tollens (GMT) - Latin for “The way that denies by denying”
- Hypothetical Syllogism

$$p \rightarrow q$$

$$\underline{p}$$

$$q$$

$$p \rightarrow q$$

$$\underline{\neg q}$$

$$\neg p$$

$$p \rightarrow q$$

$$\underline{q \rightarrow r}$$

$$p \rightarrow r$$



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p: If X is A than Y is B (Analytically known)

q: If X is A' (Analytically known)

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than Y is B' (Analytically unknown)

Where, A, B, A', B' are fuzzy terms.

A' and B' are some predicate with different linguistic hedges. To compute the membership function of B' the min – max composition of fuzzy set A' with R(x,y) which is known as implication rule is used.

$$B' = A' \circ R(x, y)$$

In terms of membership function

$$\mu_{B'}(y) = \max(\min(\mu_{A'}(x), \mu_R(x, y)))$$

where  $\mu_{B'}(y)$ ,  $\mu_{A'}(A')$ ,  $\mu_R(x, y)$  are membership function of B', A' and implication relation respectively.

p: If X is A than Y is B (Analytically known)

q: If Y is B' (Analytically known)

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than X is A' (Analytically unknown)

Where, A, B, A', B' are fuzzy terms.

$$A' = B' \circ R(x, y)$$

In terms of membership function

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# Thank You!

