

Fuzzy Sets & Expert Systems in Computer Eng. (1):

Fuzzy Sets

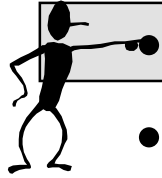
Piero P. Bonissone

GE Corporate Research & Development

Bonissone@crd.ge.com

**(adapted from slides by Roger Jang
and Enrique Ruspini)**

Outline



Motivation

- **Fuzzy Sets Basic Concepts**

- **Characteristic Function (Membership Function)**
- **Examples**
- **Notation**
- **Semantics and Interpretations**
- **Related crisp sets**
 - » **Support, Bandwidth, Core, α -level cut**
 - » **Decomposition Theorem**
- **Features, Properties, and More Definitions**
 - » **Convexity, Normality**
 - » **Cardinality, Measure of Fuzziness, First Moment**
 - » **MF parametric formulation**

- **Fuzzy Set-theoretic Operations**

- **Intersection, Union, Complementation**
- **Numerical Examples**
- **T-norms and T-conorms**

Energy Strategy Draws Congressional Criticism

... the Administration's national energy policy was criticized today as one sided even by those who *vigorously support* its plan for drilling in *environmentally fragile* areas.

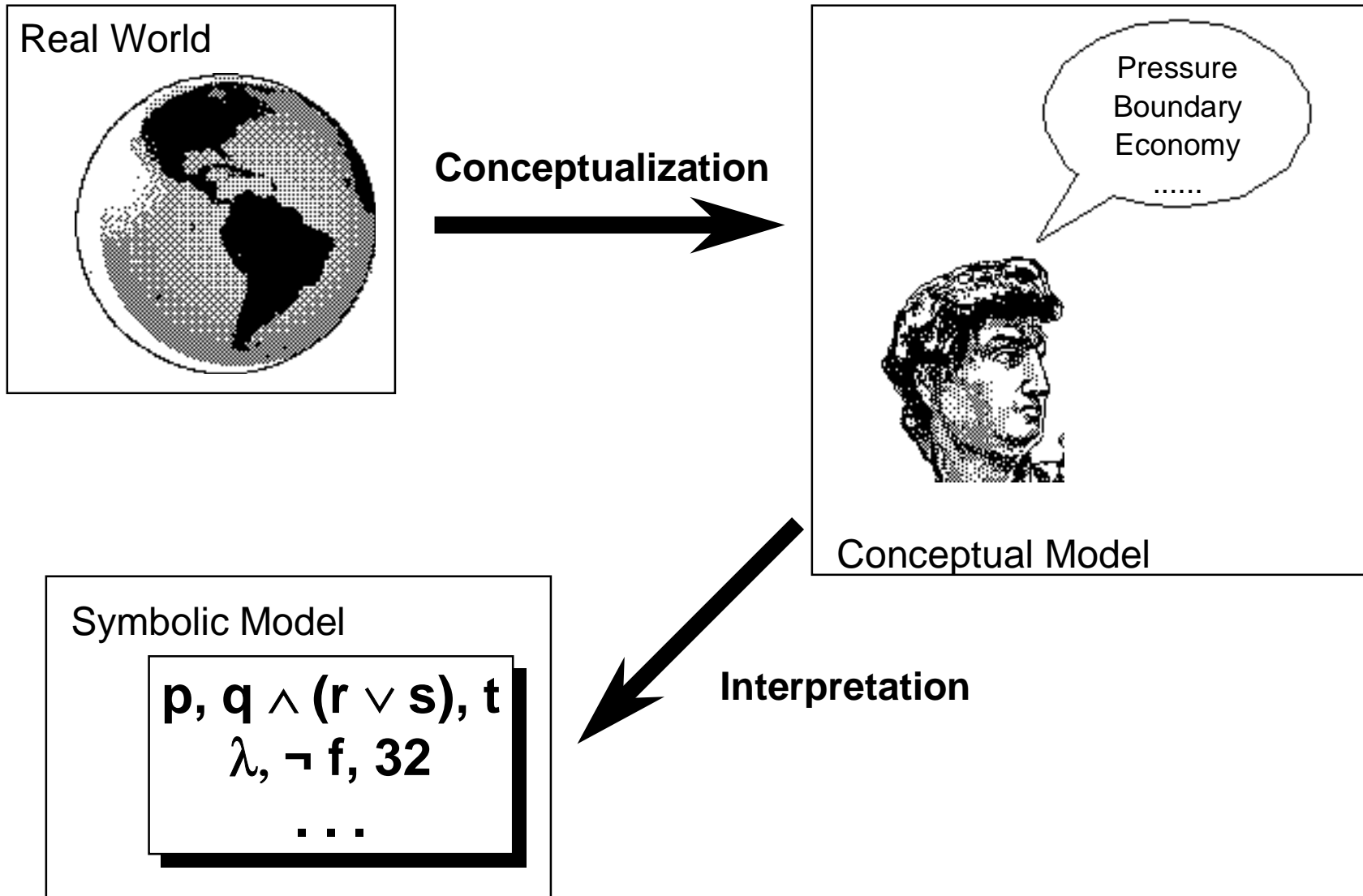
... Mr. Johnston, a *strong supporter* of more oil and gas drilling, as well as others forms of energy, said “You've just got to do that for *balance*; I don't care if you *believe* it or not.”

New York City Population Gain Attributed to Immigrant Tide

The white population of Manhattan, including some Hispanic residents, grew by 20,263 during the decade, to 867,227, a gain of 3.1 percent. Staten Island added 8509 whites for a total of 322,043, a gain of 2.7 percent.

(New York Times, February 22, 1991)

MODELING



Approximate Models

Imprecise, Vague, Uncertain Representations of System Behavior

- *Classical Models:*

If Pressure = 10 ATM, then Volume = 2.5 Cm

- *Imprecise Models:*

If Pressure \geq 5 ATM, then Volume \leq 6 Cm

- *Uncertain Models:*

If Pressure \geq 5 ATM, then Prob(Vol = 6 Cm) = 0.9

- *Vague Models:*

If Pressure is HIGH, then Volume is LOW

Approximate Models and Decision/Control Rules

- Vague rules may be used to describe characteristics of the system:

*If Position(t) is NEAR and Velocity is HIGH, then
Position(t+1) is MEDIUM,*

*If Shape is ROUND and Gap is SMALL, then
Probability(Symbol=a) is HIGH,*

- Usually, the problem-solving goal is the generation of decision and control rules:

*If Position(t) is LOW and Velocity(t) is HIGH, then
the Acceleration should be SMALL,*

*If Probability(Symbol1=a) is LOW and
Probability(Symbol2=x) is HIGH, then
Probability(Sequence=ex) is HIGH.*

What is Fuzzy Logic ?

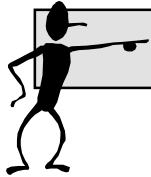
- Inferential Approach
- Oriented towards System Analysis/Decision Support
- Utilized to develop Intelligent Automated Systems
- Capable of dealing with Vague Information
- Facilitates development of qualitative models
- Extension of Classical Logic using Multiple Truth-Values
- Exploits notions of similarity between situations
- Based on the Theory of Fuzzy Sets (Zadeh 1965)

Why Fuzzy Logic ?

- Ability to translate imprecise/vague knowledge of human experts
- Simple, easy to implement technology
- Rules contribute to inferences even when facts do not exactly match antecedent
- Rule-based systems may be analyzed and improved
- Technology is easy to transfer from product to product
- Smooth controller behavior
- Robust controller behavior
- Ability to control unstable systems

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Fuzzy Sets

$$A = \{x \in X \mid x > 10\}$$

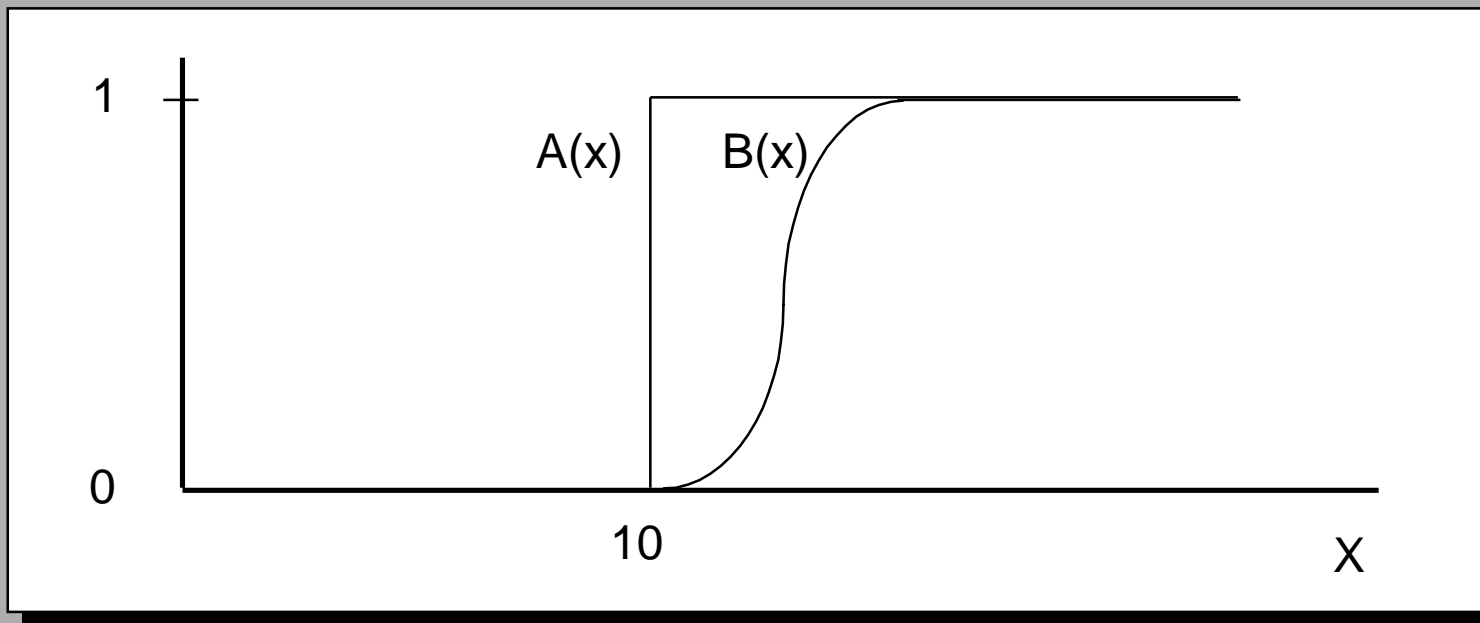
$$B = \{x \in X \mid x \gg 10\}$$

Boolean Set

Fuzzy Set

$$A: X \rightarrow \{0, 1\}$$

$$B: X \rightarrow [0, 1]$$



Characteristic function of sets A(x) and B(x)

Boolean Algebra (Run through)

Assign binary truth value to statements

| A | statement |
|---|-----------|
| 1 | true |
| 0 | false |

| A | $\neg A$ |
|---|----------|
| 1 | 0 |
| 0 | 1 |

Combine statements using AND and OR operators

| A | B | $A \vee B$ |
|---|---|------------|
| 0 | 0 | 0 |
| 0 | 1 | 1 |
| 1 | 0 | 1 |
| 1 | 1 | 1 |

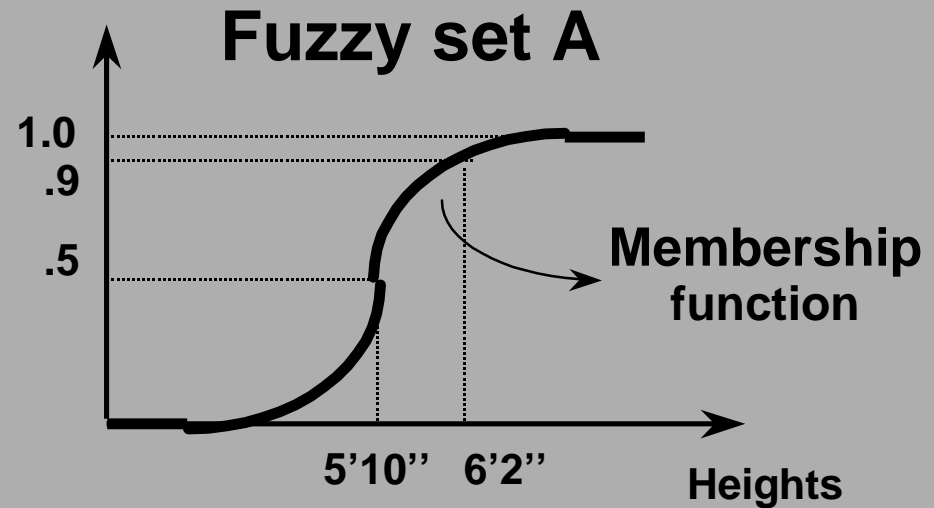
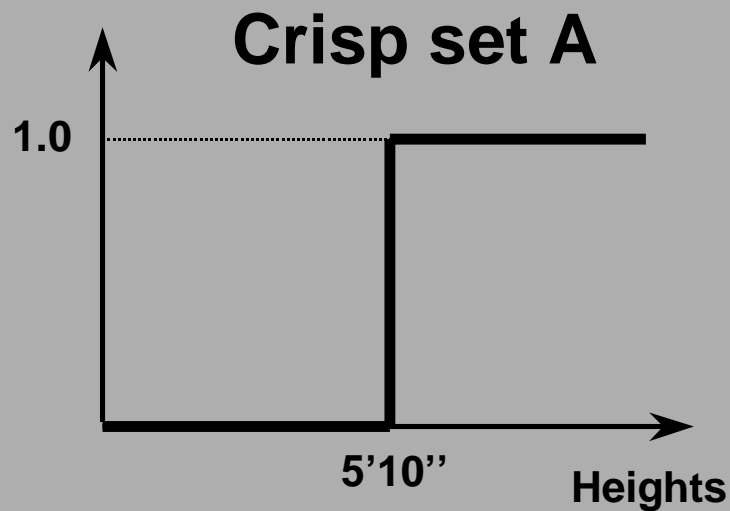
| A | B | $A \wedge B$ |
|---|---|--------------|
| 0 | 0 | 0 |
| 0 | 1 | 0 |
| 1 | 0 | 0 |
| 1 | 1 | 1 |

Fuzzy Sets

Binary Logic vs. Fuzzy Logic:

Sets with crisp and fuzzy boundaries, respectively

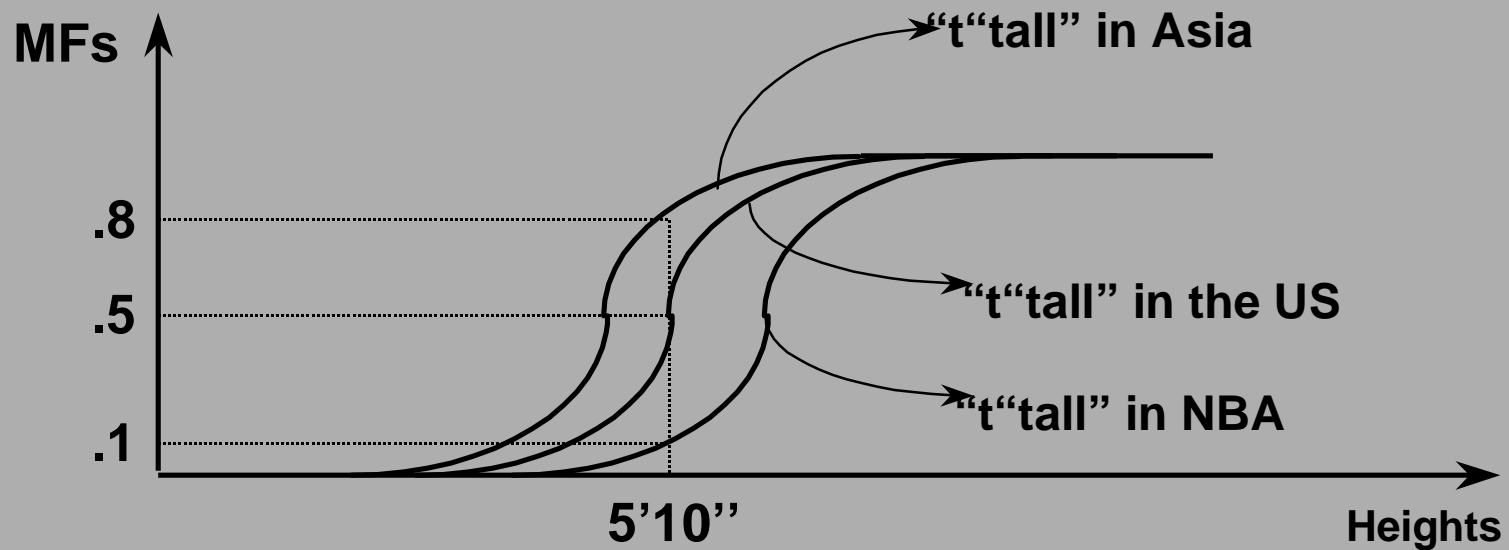
A = Set of tall people



Membership Functions (MFs)

Characteristics of MFs:

- Subjective measures
- Not probability functions



Fuzzy Sets

Formal definition:

A fuzzy set A in X is expressed as a set of ordered pairs:

$$A = \{(x, \mu_A(x)) \mid x \in X, \mu_A(x): X \mapsto [0,1]\}$$

Fuzzy set

Membership
function
(MF)

Universe or
universe of discourse

A fuzzy set is totally characterized by a membership function (MF).

Fuzzy Sets with Discrete Universes

Fuzzy set C = “desirable city to live in”

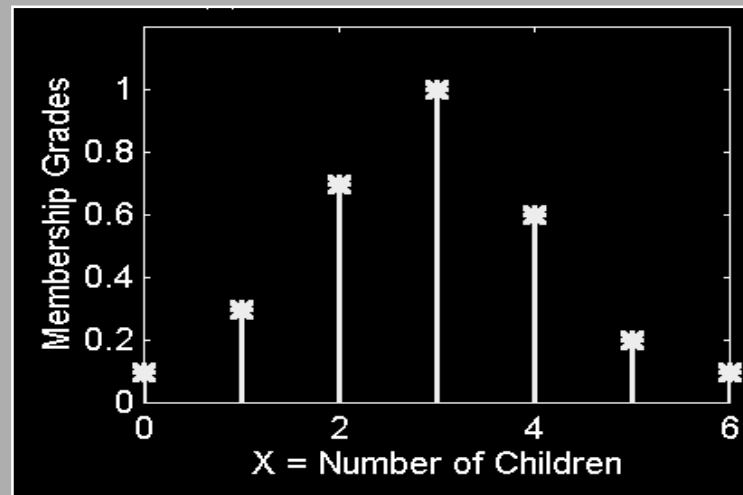
$X = \{\text{SF, Boston, Troy}\}$ (discrete and nonordered)

$C = \{(\text{SF}, 0.9), (\text{Boston}, 0.8), (\text{Troy}, 0.6)\}$

Fuzzy set A = “sensible number of children”

$X = \{0, 1, 2, 3, 4, 5, 6\}$ (discrete universe)

$A = \{(0, .1), (1, .3), (2, .7), (3, 1), (4, .6), (5, .2), (6, .1)\}$



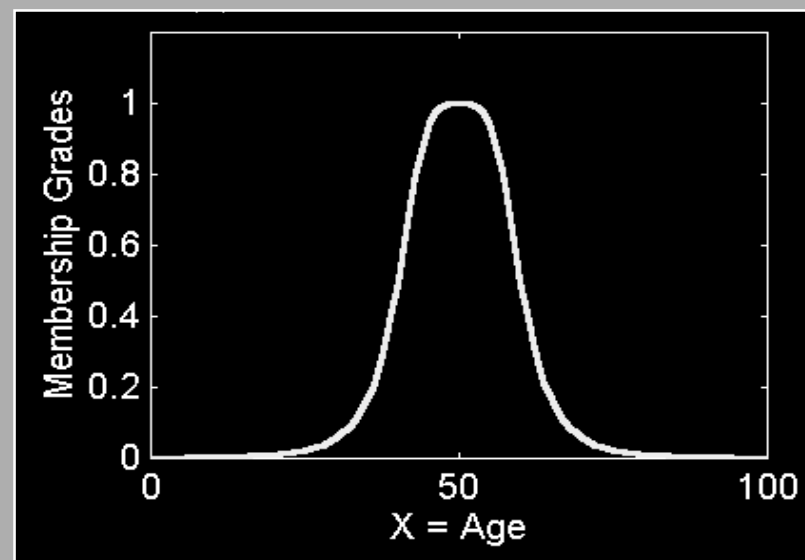
Fuzzy Sets with Cont. Universes

Fuzzy set B = “about 50 years old”

X = Set of positive real numbers (continuous)

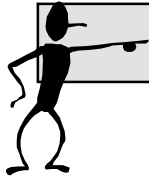
B = {(x, $\mu_B(x)$) | x in X}

$$\mu_B(x) = \frac{1}{1 + \left(\frac{x - 50}{10}\right)^2}$$



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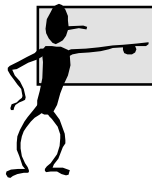


Examples of Fuzzy Sets

- Tall Persons (*Height*)
- Dangerous Maneuvers (*Action Sequences*)
- Blonde Individuals (*Hair color*)
- Loud Noises (*Sound Intensity*)
- Large Investments (*Money*)
- High Speeds (*Speed*)
- Close Objects (*Distance*)
- Large Numbers (*Numbers*)
- Desirable Actions (*Decision or Control Space*)

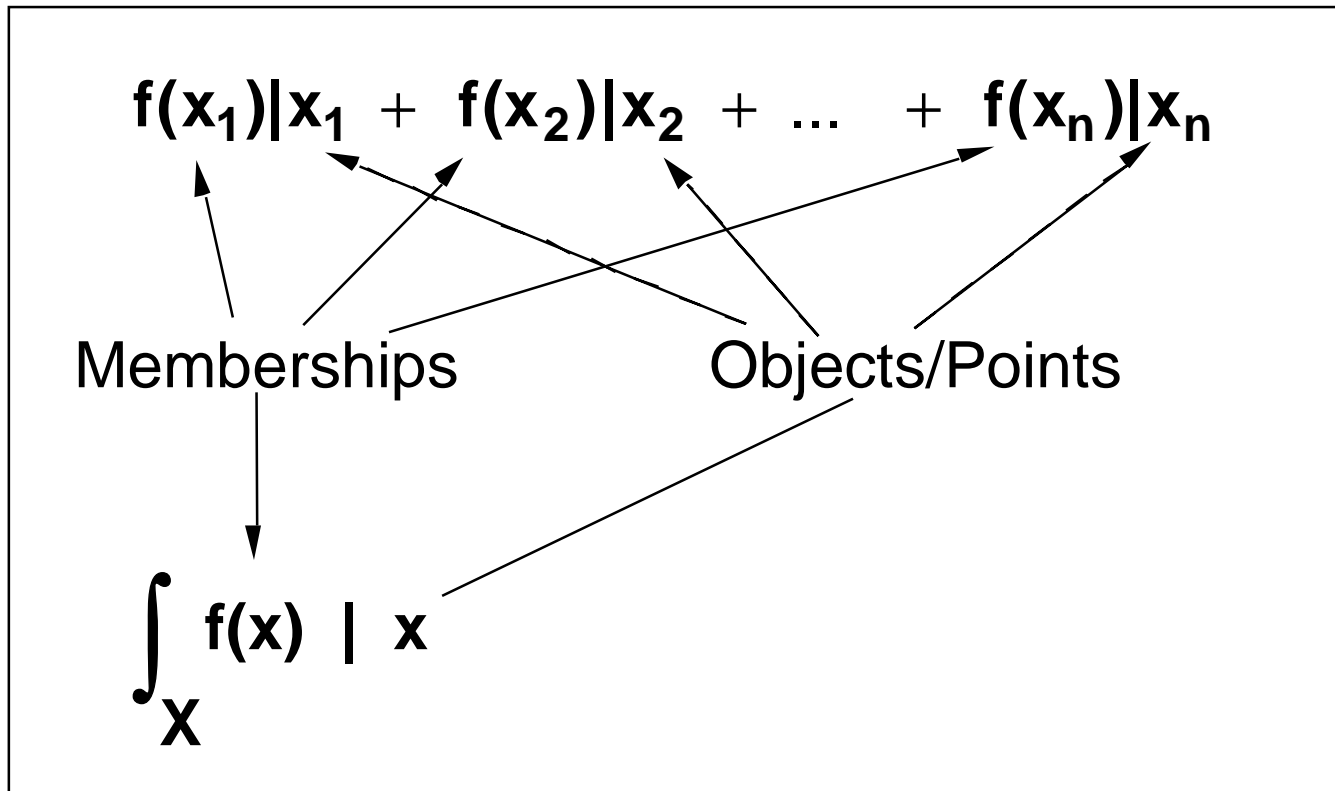
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Denoting Fuzzy Sets

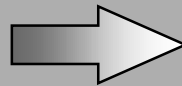
$$f : X \rightarrow [0,1] : x \rightarrow f(x)$$



Alternative Notation

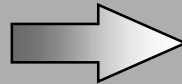
A fuzzy set A can be alternatively denoted as follows:

X is discrete



$$A = \sum_{x_i \in X} \mu_A(x_i) / x_i$$

X is continuous

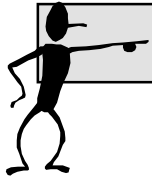


$$A = \int_X \mu_A(x) / x$$

Note that Σ and integral signs stand for the union of membership grades; “/” stands for a marker and does not imply division.

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Probability vs. Fuzziness

Randomness:

Uncertainty described by tendency (frequency) of a random variable to take on a value in a specified region

Interpretations: frequency -> willingness to accept bet (subjective probability)

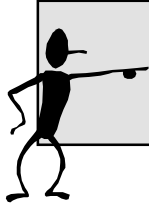
Fuzziness:

Degree to which the element satisfies properties characterized by a fuzzy set.

Interpretations: Possibility -> similarity -> desirability

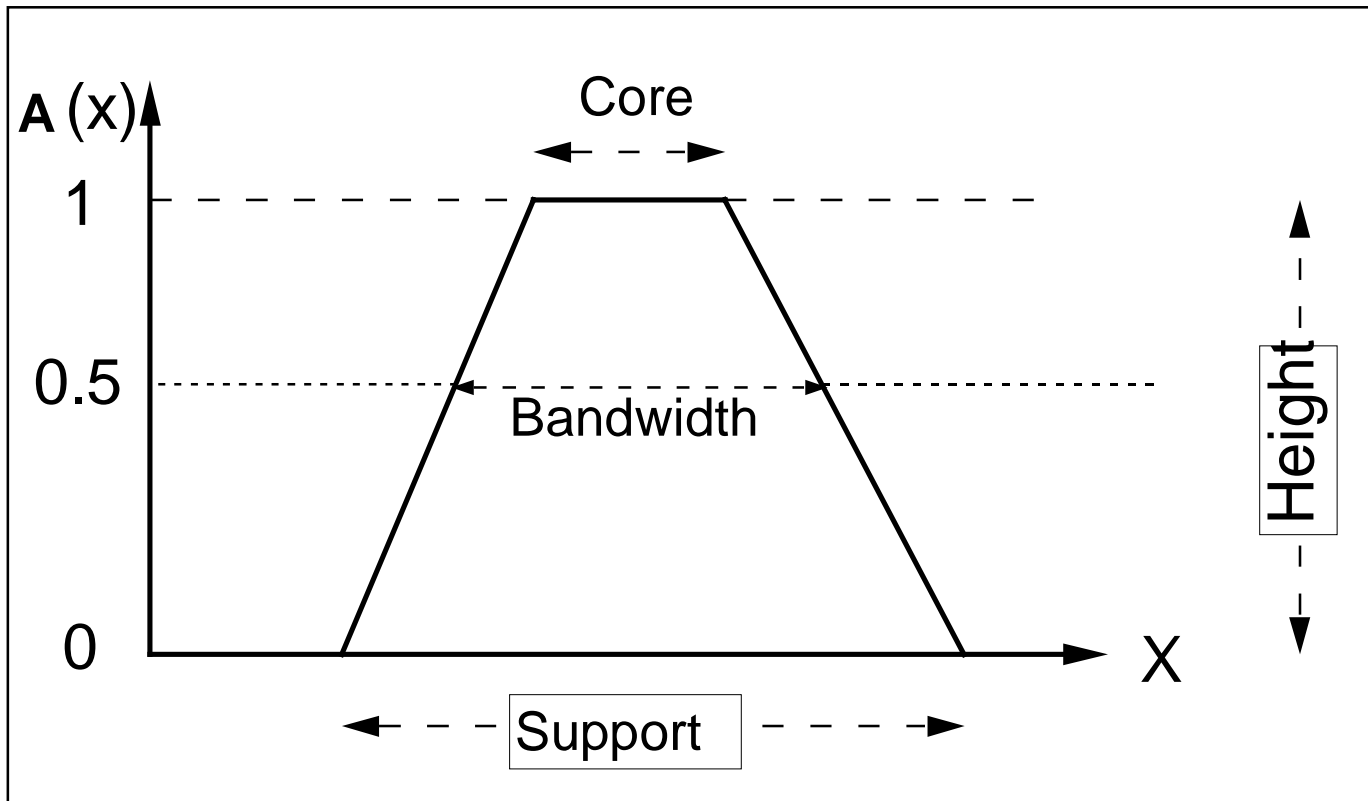
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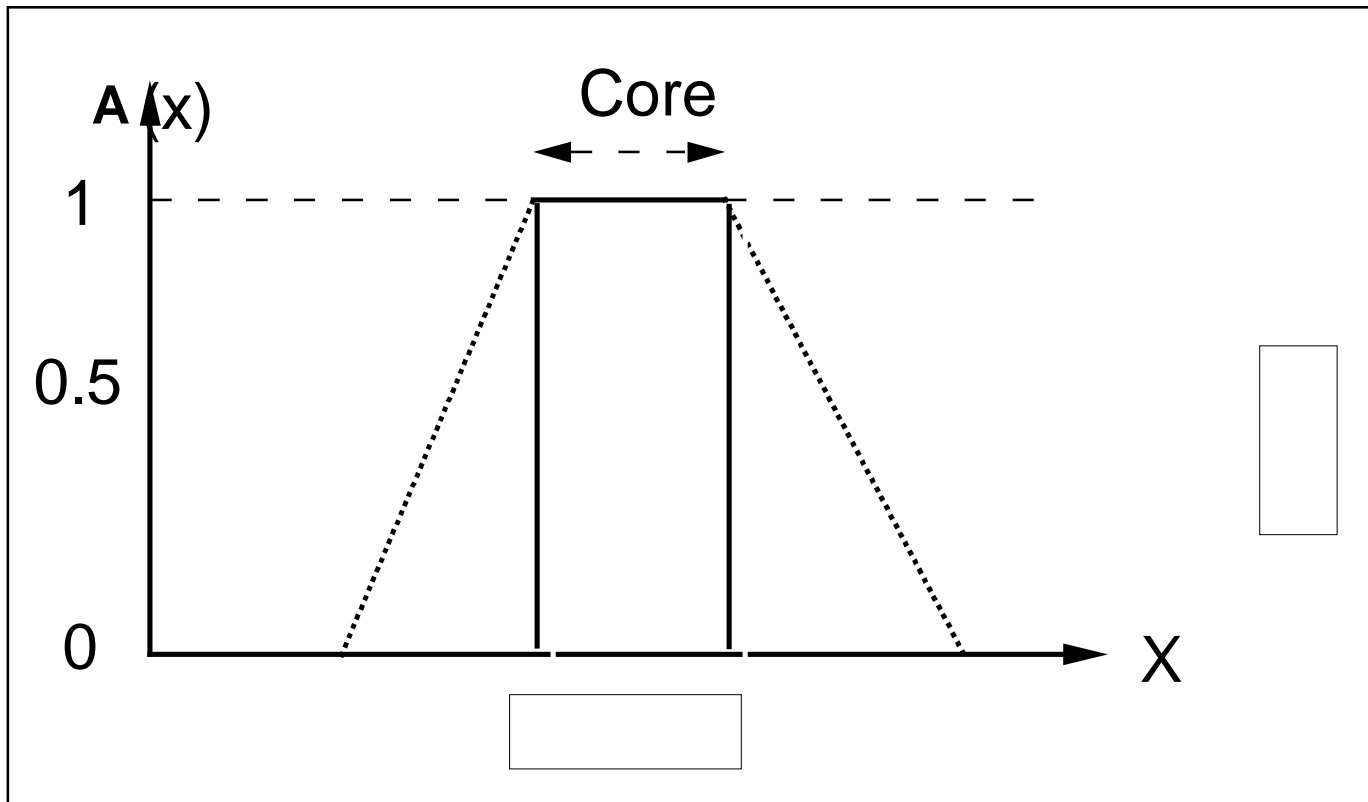
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Fuzzy Sets



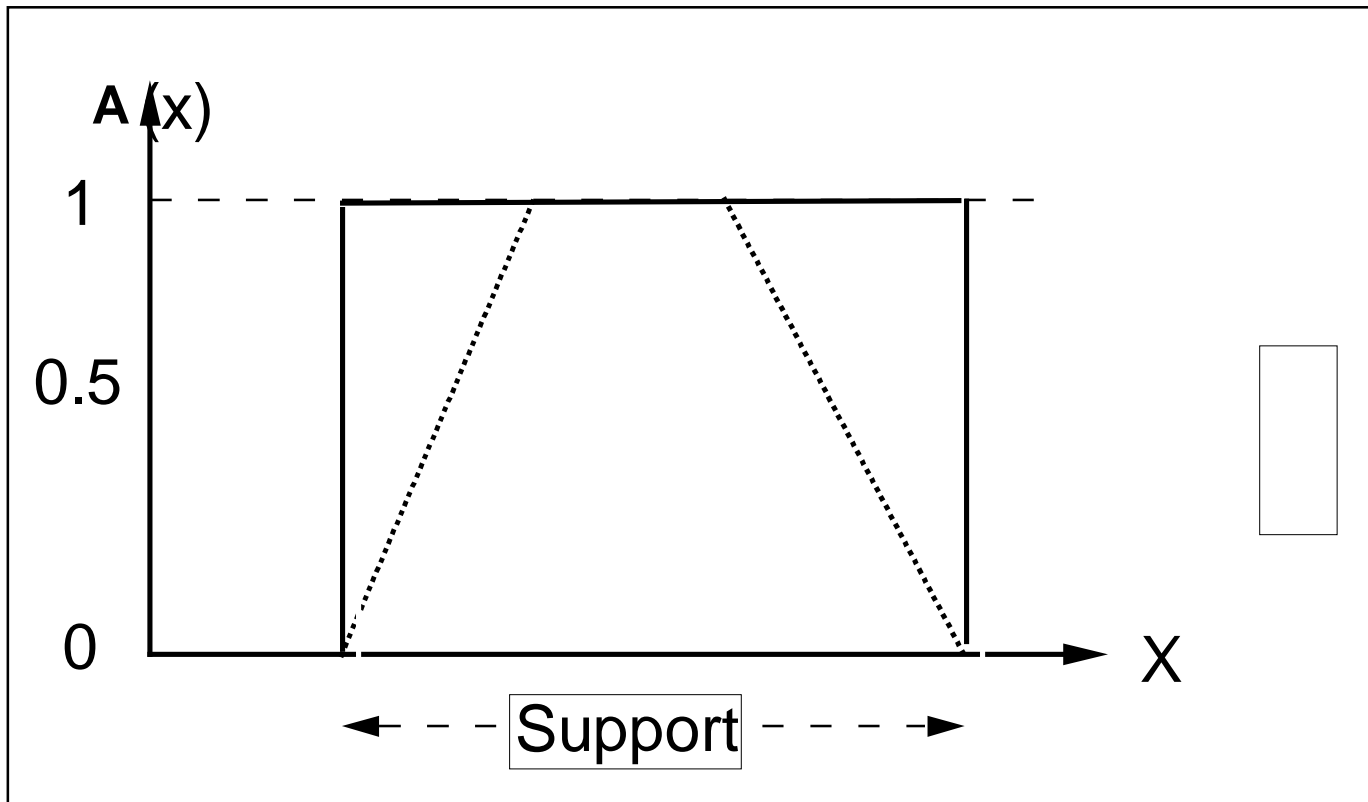
$$\begin{aligned}\text{Core}(A) &= \{x \mid A(x) = 1\} \\ \text{Support}(A) &= \{x \mid A(x) > 0\}\end{aligned}$$

Fuzzy Sets



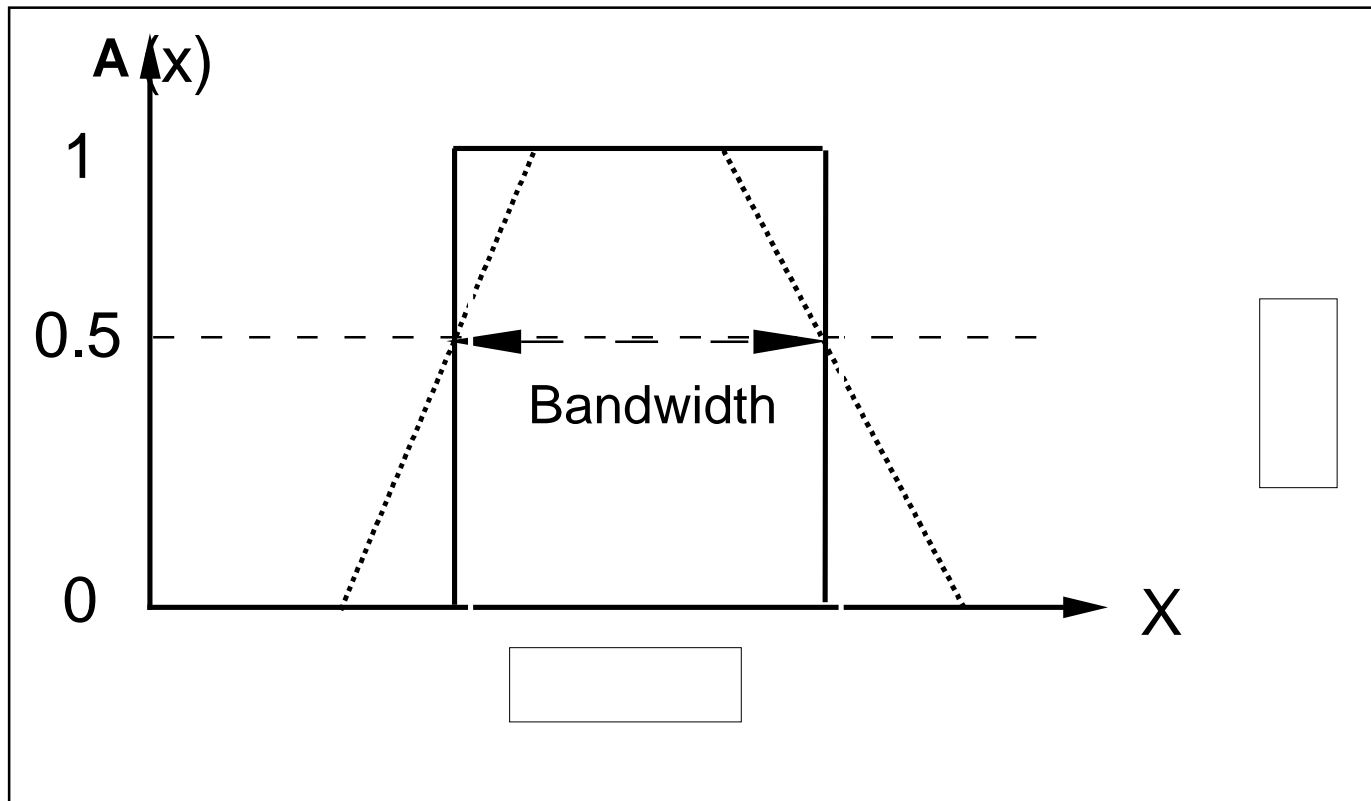
$$\text{Core}(A) = \{x \mid A(x) = 1\}$$

Fuzzy Sets



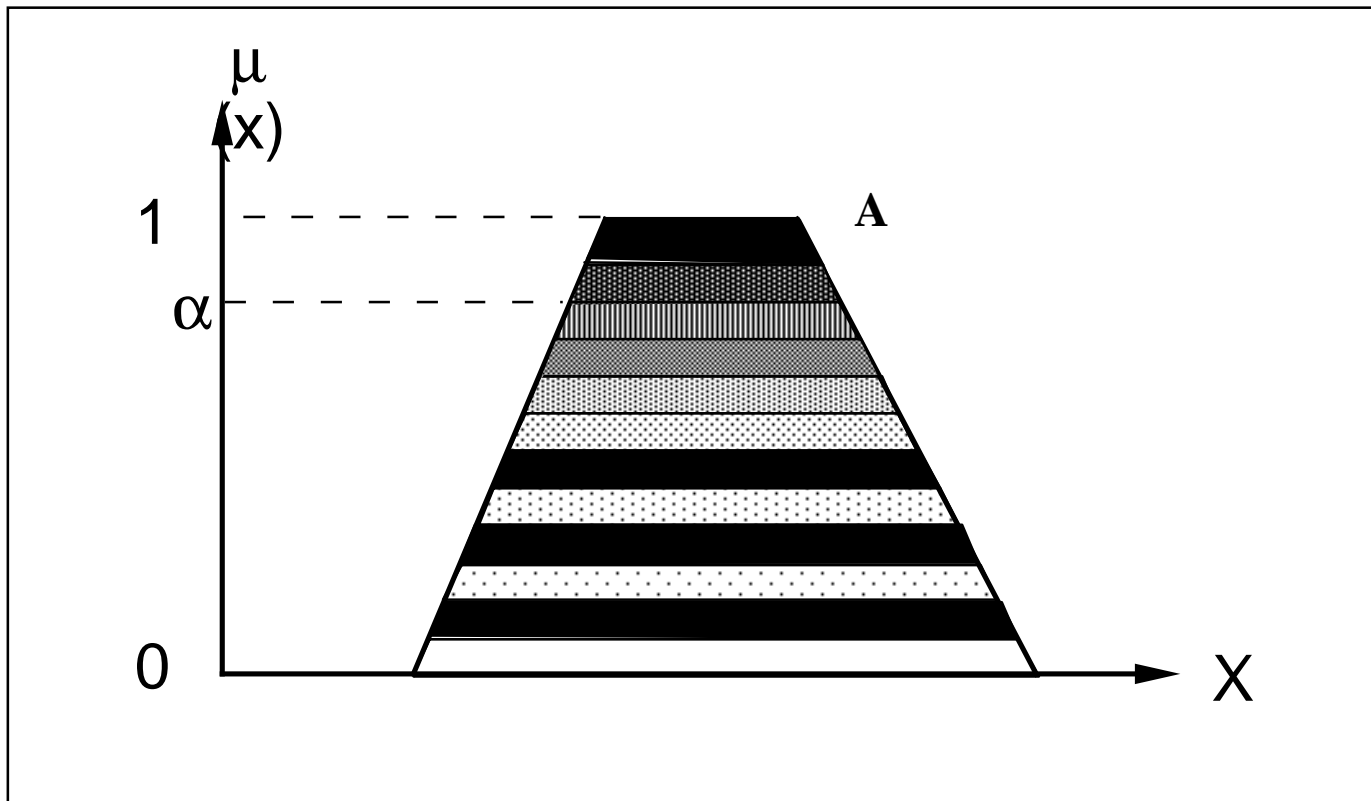
$$\text{Support}(A) = \{x \mid A(x) > 0\}$$

Fuzzy Sets



$$\text{Bandwidth}(A) = \{x \mid A(x) \geq 0.5\}$$

Fuzzy Sets as Collections of Conventional Sets



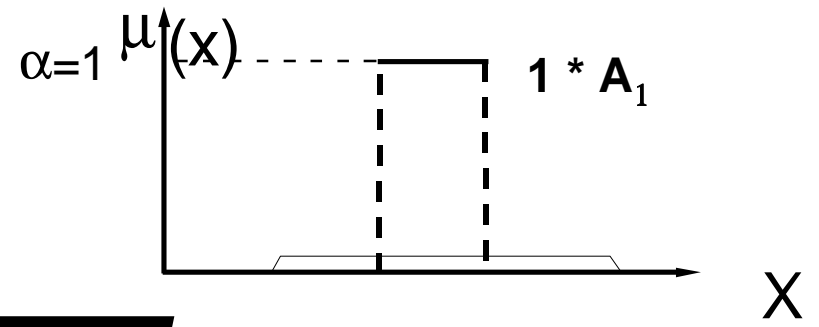
$$A_{\alpha} = \{x \mid A(x) \geq \alpha\},$$

Identity Principle (Decomposition Theorem)

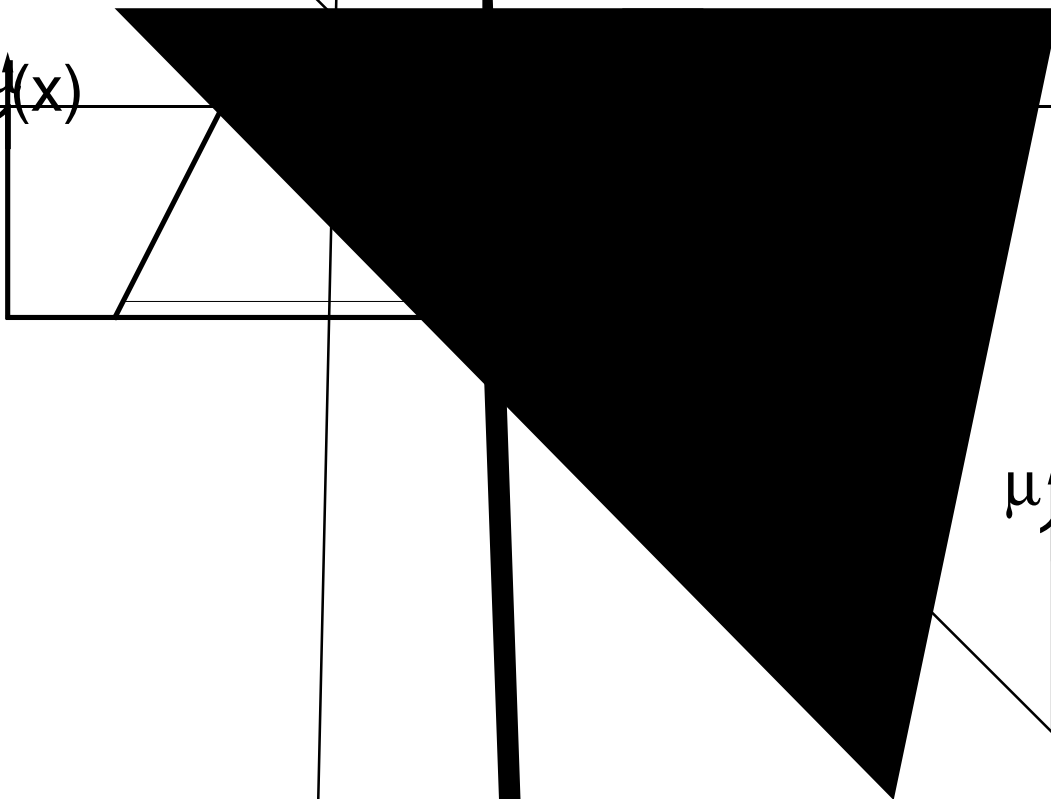
- A fuzzy set A can be represented by the union of all its α -cut sets, weighted by their value α :

$$A(x) = \bigcup_{\alpha \in [0,1]} \alpha \cdot A_{\alpha}(X)$$

Identity Principle (Decomposition Theorem)

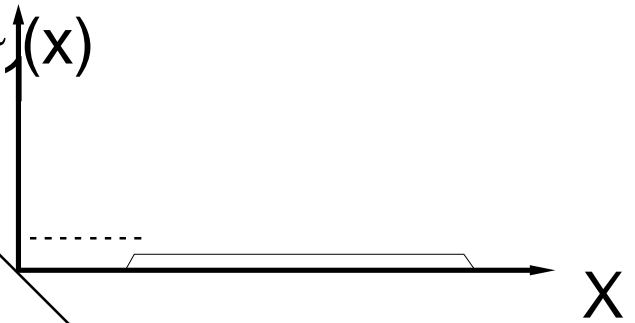


$\mu(x)$

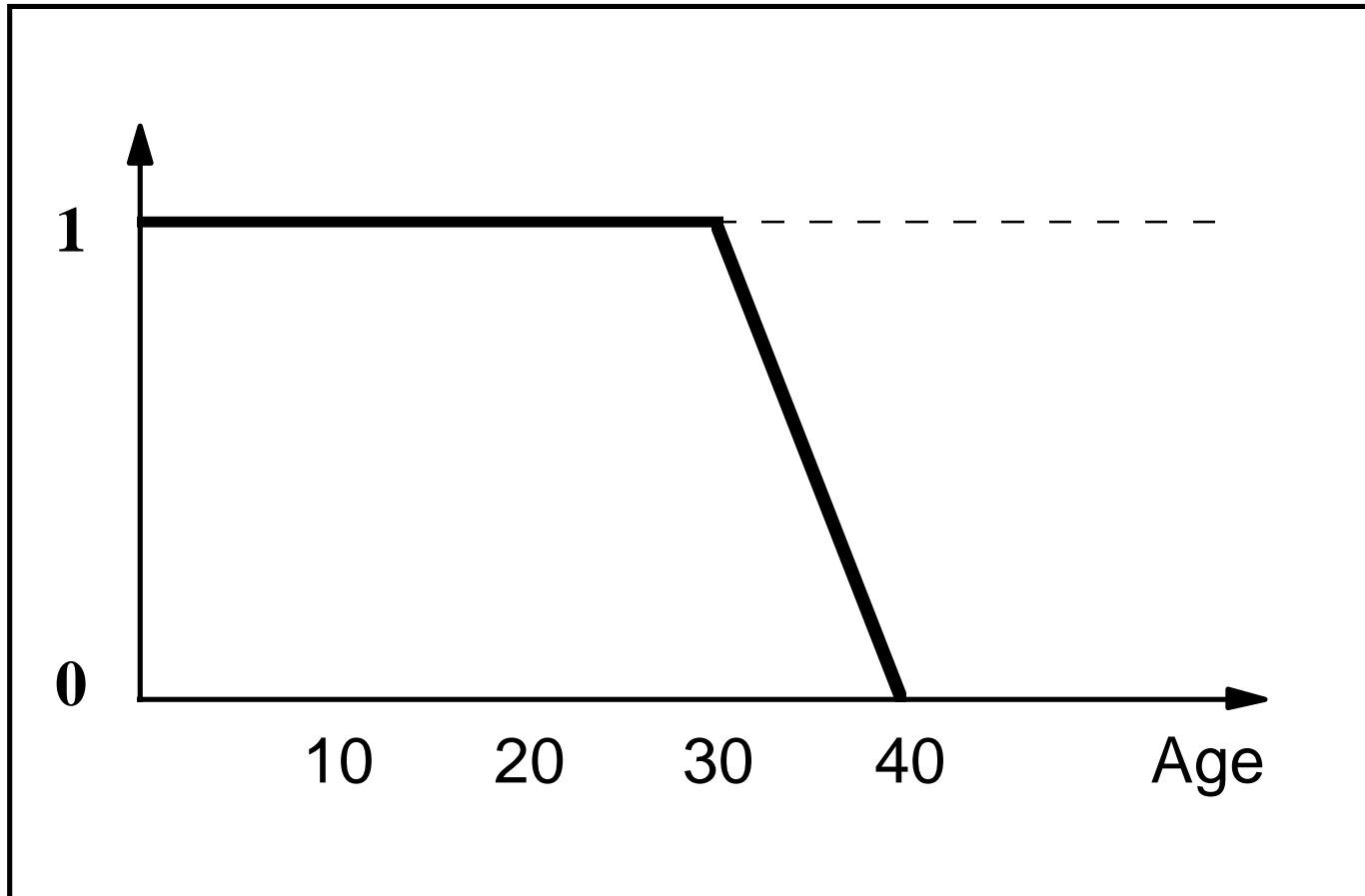


μ

$\mu_j(x)$

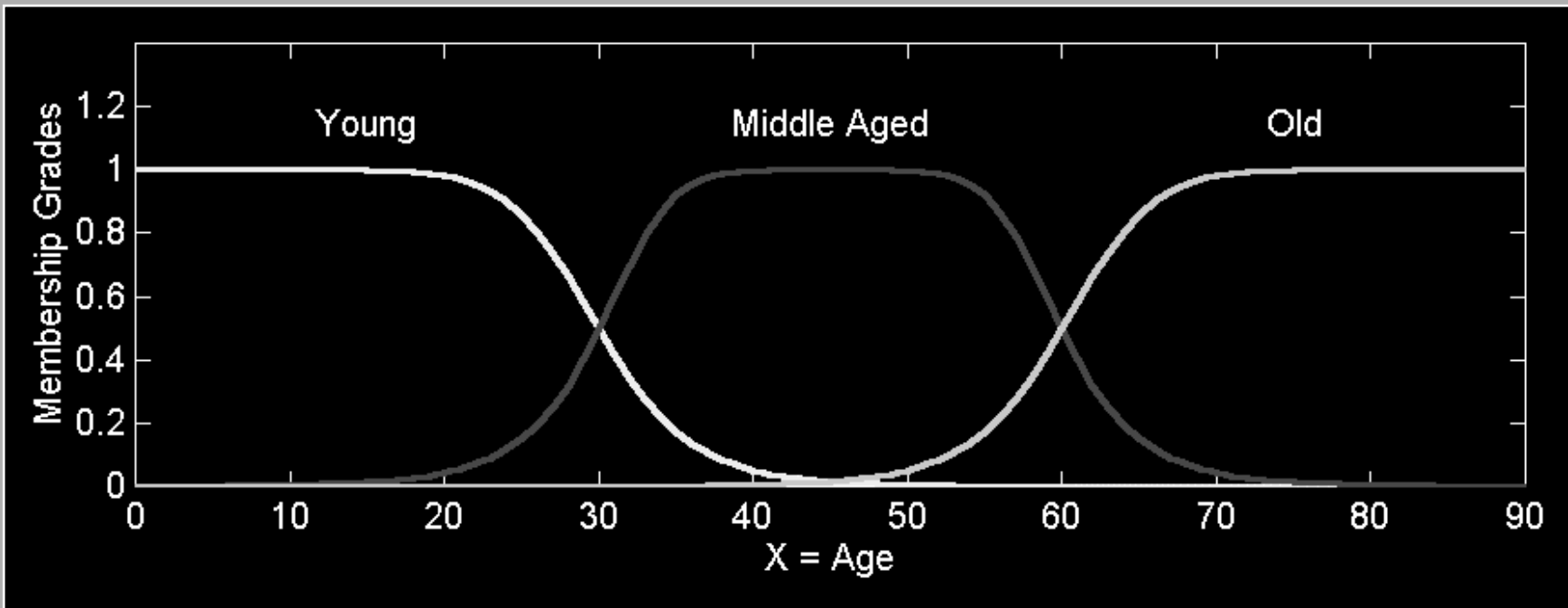


Young Persons



Fuzzy Partition

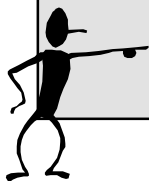
Fuzzy partitions formed by the linguistic values “young”, “middle aged”, and “old”:



lingmf.m

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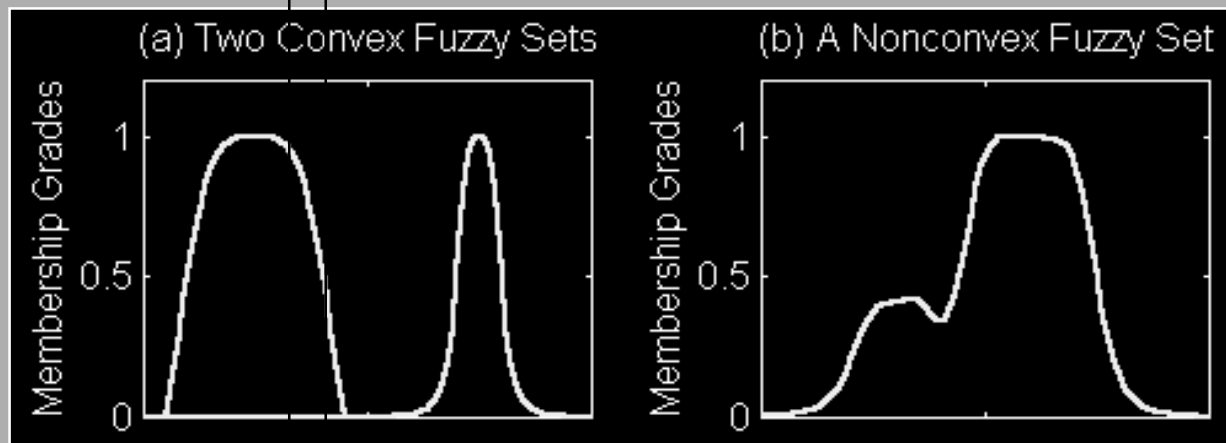


Convexity of Fuzzy Sets

A fuzzy set A is convex if for any λ in $[0, 1]$,

$$\mu_A(\lambda x_1 + (1 - \lambda)x_2) \geq \min(\mu_A(x_1), \mu_A(x_2))$$

Alternatively, A is convex if all its α -cuts are convex.

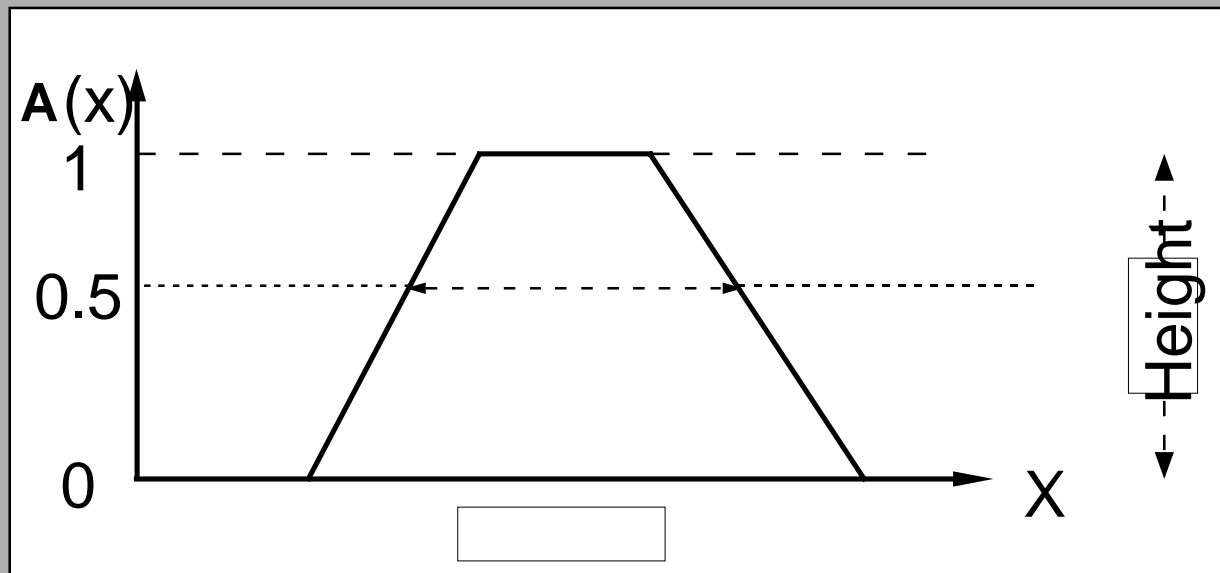


convexmf.m

Normality of Fuzzy Sets

A fuzzy set A is *normal* if

$$\text{Height}(A) = \text{Max}_x A(x) = 1$$



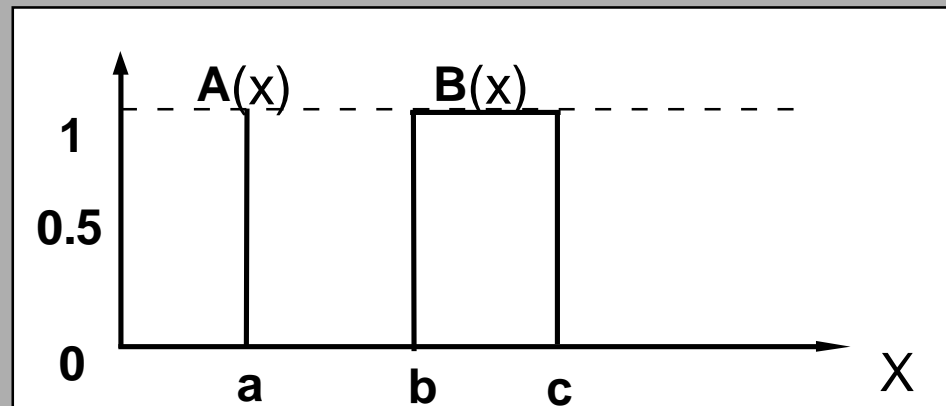
Fuzzy Set Representation of Crisp Numbers and Crisp Intervals

A crisp number a is represented by a fuzzy singleton

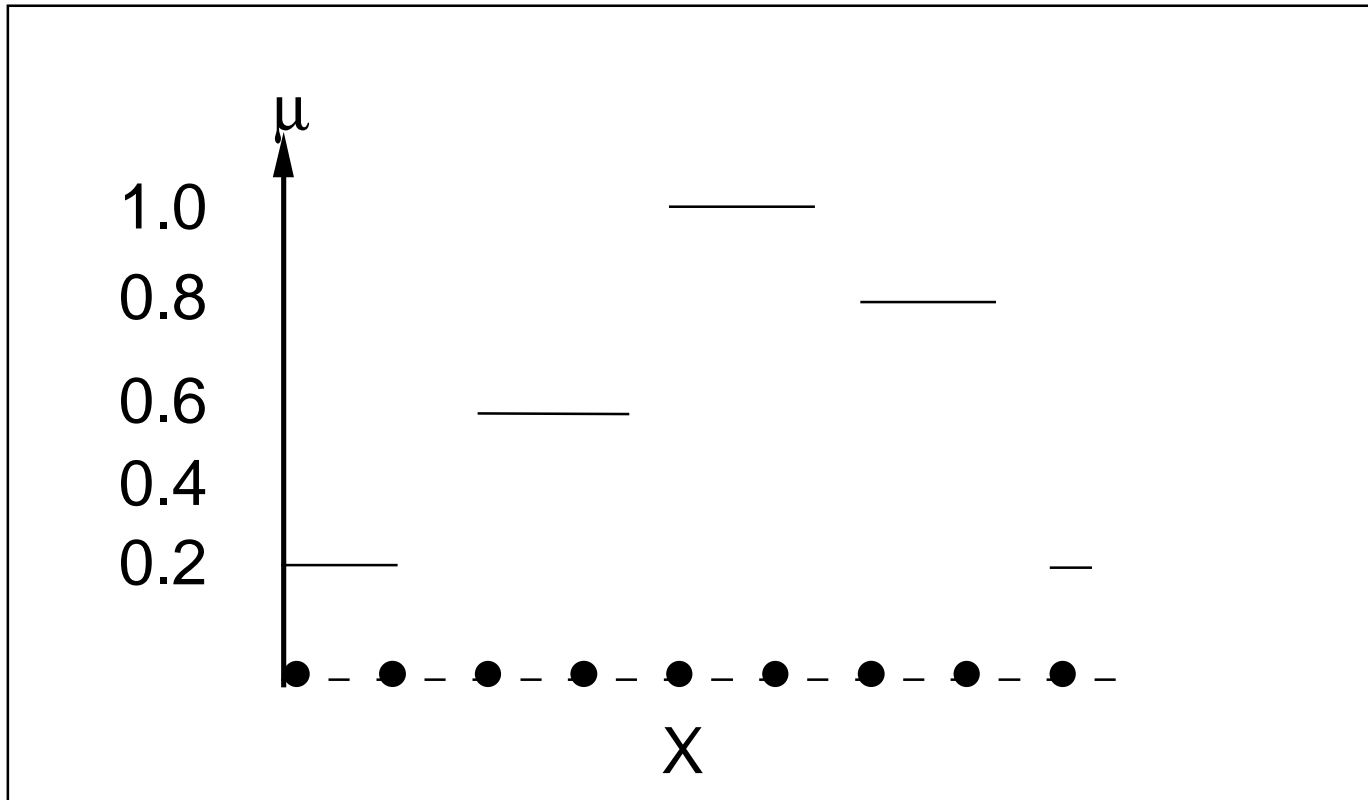
$$A(x) = \begin{cases} 1 & \text{if } x = a \\ 0 & \text{if } x \neq a \end{cases}$$

A crisp interval $[b,c]$ is represented by a fuzzy set

$$B(x) = \begin{cases} 1 & \text{if } x \in [b, c] \\ 0 & \text{if } x \notin [b, c] \end{cases}$$

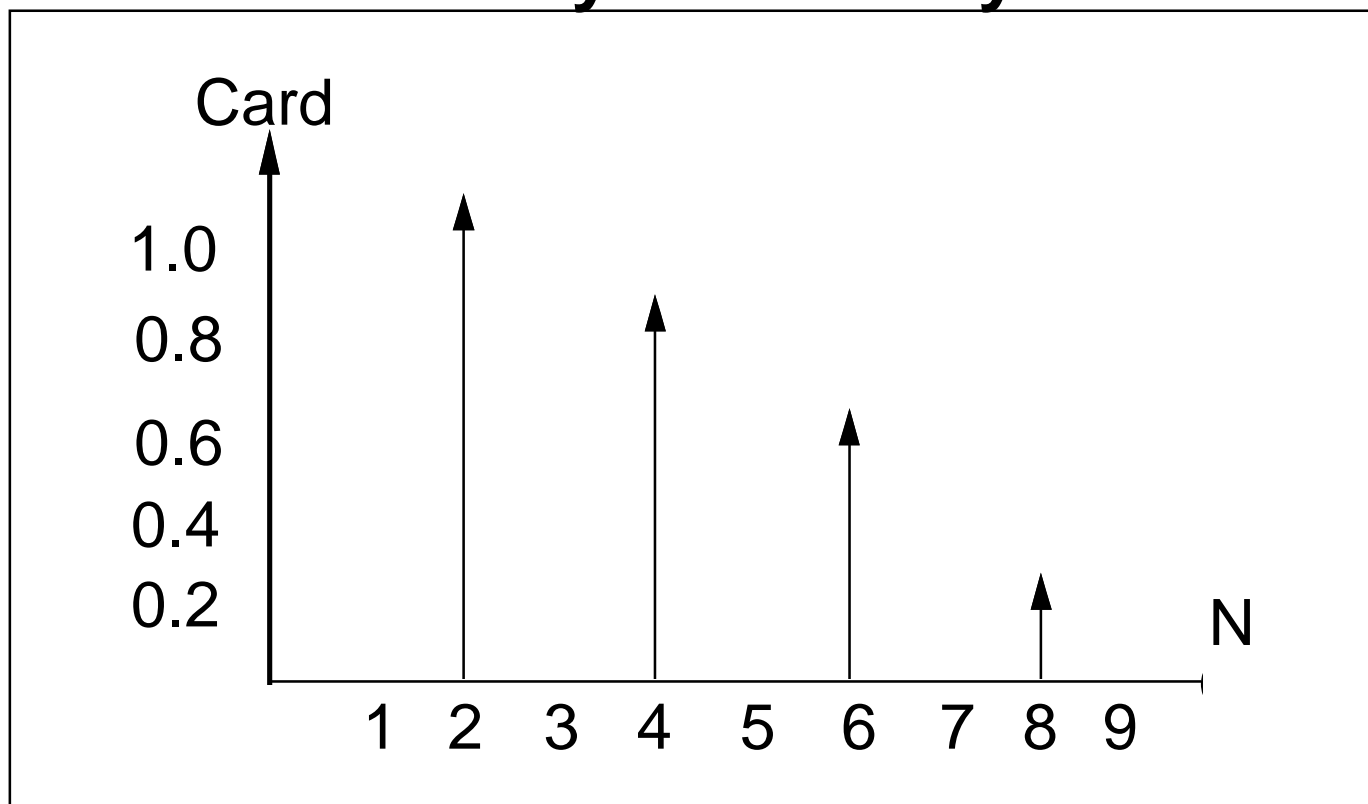


Scalar Cardinality



$$\Sigma \text{ Count (A)} = \text{Card(A)} = 5.4$$

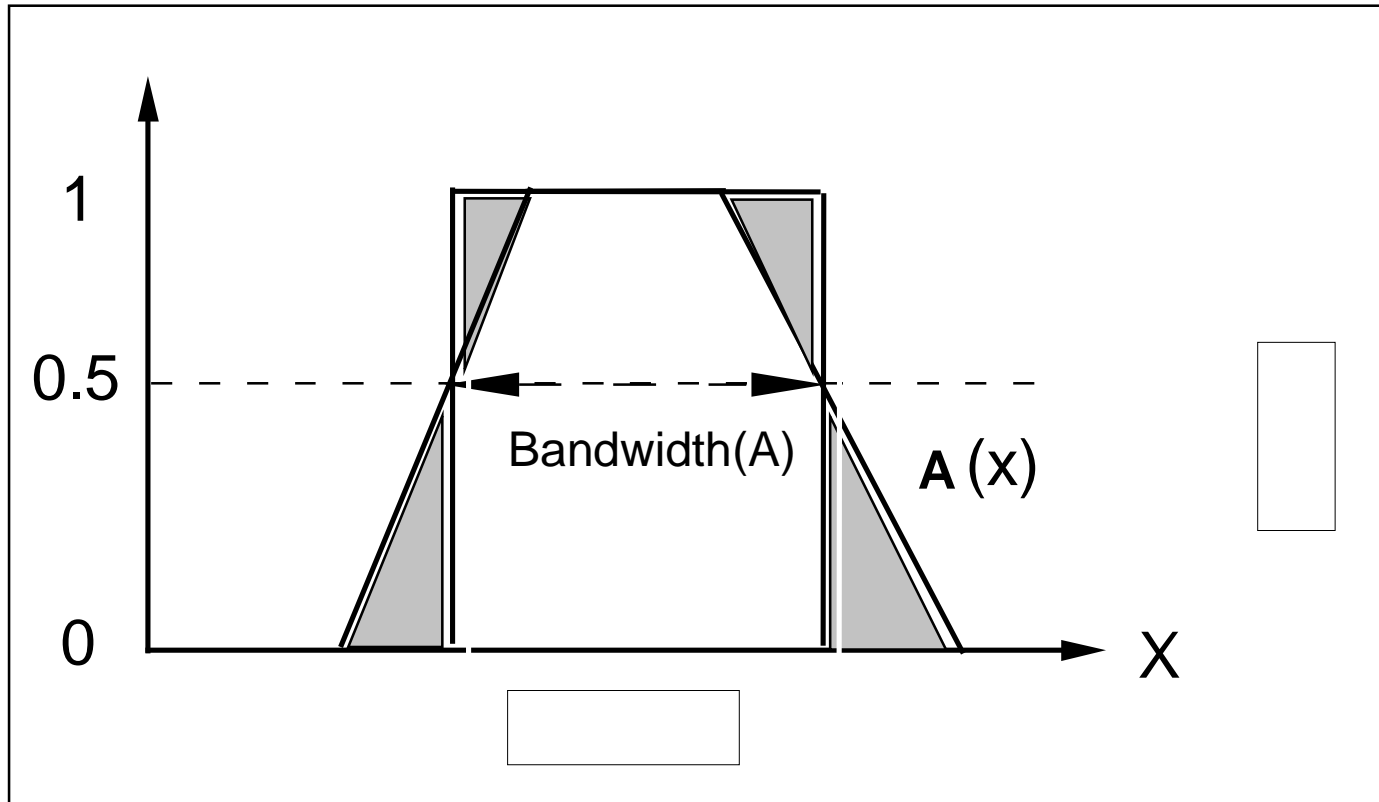
Fuzzy Cardinality



$$|A|(n) = \begin{cases} \alpha, & \text{if exists an } \alpha\text{-cut with } |A_\alpha| = n, \\ 0, & \text{otherwise.} \end{cases}$$

[The answer is a *fuzzy set* in the set of integer numbers]

Measure of Fuzziness



**Measure of Fuzziness = Cardinality $\{|Bandwidth(A) - A(x)|\}$
= Cardinality $\{\text{■}\}$**

First Moment of a Fuzzy Sets

The First Moment of a discrete Fuzzy Set $A(x_i)$ is:

$$FirstMoment(A) = \frac{\sum_{i=1}^n A(x_i) * x_i}{\sum_{i=1}^n A(x_i)}$$

The First Moment of a continuous Fuzzy Set $A(x)$ is:

$$FirstMoment(A) = \frac{\int A(x) * x dx}{\int A(x) dx}$$