



Institut für Industrielle
Informationstechnik
Institute Industrial IT






Information Fusion – Basics on ET

Combination Techniques for Uncertain Information in Measurement and Signal Processing

3.3 Fuzzy Set Theory

Information Fusion

Prof. Dr.-Ing. Volker Lohweg



Institut für Industrielle
Informationstechnik
Institute Industrial IT

Lectures – Contents

<p>1. Introduction</p> <ul style="list-style-type: none"> 1.1 Why Information Fusion? 1.2 Information and Measurement Taxonomy of Uncertainty 1.3 Information and Pattern Recognition <p>2. Basics on Information Fusion</p> <ul style="list-style-type: none"> 2.1 Basics 2.2 Concepts 2.3 Strategies 	<p>3. Basics on Evidence Theory</p> <ul style="list-style-type: none"> 3.1 Probability Theory 3.2 Dempster-Shafer Theory 3.3 Fuzzy Set Theory 3.4 Possibility Theory <p>4. Inference and Fusion Strategies</p> <ul style="list-style-type: none"> 4.1 Evidence Theory Fusion 4.2 Interconnections <p>5. Fusion: Research and Application</p> <ul style="list-style-type: none"> 5.1 New Research Concepts 5.2 Examples
--	---

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

2

inTT Institut für Industrielle Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Introduction

- Example

Precision and Significance in the Real World

Precision

Significance

Information Fusion Only valid with verbal explanations Prof. Dr.-Ing. Volker Lohweg 3

inTT Institut für Industrielle Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Introduction

- "Many decision-making and problem-solving tasks are **too complex** to be understood quantitatively, however, people succeed by using knowledge which is imprecise rather than precise."
- *Fuzzy Set Theory*, originally introduced by Lotfi Zadeh in the 1960's, resembles human reasoning in its use of approximate information and uncertainty to generate decisions.
- It was specifically designed to mathematically represent *uncertainty and vagueness* and provide formalized tools for dealing with the *imprecision* intrinsic to many problems. By contrast, traditional computing demands precision down to each bit.

Information Fusion Only valid with verbal explanations Prof. Dr.-Ing. Volker Lohweg 4

3.3 Fuzzy Set Theory – Introduction


- Fuzzy, or multi-valued logic was introduced in the 1930s by **Jan Lukasiewicz**, a Polish philosopher.
- While classical logic operates with only two values 1 (true) and 0 (false), Lukasiewicz introduced logic that extended the range of truth values to all real numbers in the interval between 0 and 1.
- He used a number in this interval to represent the **possibility** that a given statement was true or false. For example, the possibility that a man 181 cm tall is really tall might be set to a value of 0.86.
- It is *likely* that the man is tall. This work led to an inexact reasoning technique often called **Possibility Theory** coined by Zadeh and Dubois & Prade.




3.3 Fuzzy Set Theory – Introduction

- Later, in 1937, **Max Black** published a paper "Vagueness: an exercise in logical analysis".
- In this paper, he argued that a **continuum** implies degrees. Imagine, he said, a line of countless "chairs". At one end is a Chippendale. Next to it is a near-Chippendale, in fact indistinguishable from the first item. Succeeding "chairs" are less and less chair-like, until the line ends with a log.
- When does a *chair* become a *log*? Max Black stated that if a continuum is discrete, a number can be allocated to each element.
- He accepted **vagueness as a matter of probability**.






Institut für Industrielle
Informationstechnik
Institute Industrial IT



3.3 Fuzzy Set Theory – Introduction

- In 1965 **Lotfi Zadeh**, published his famous paper "Fuzzy Sets".
- Zadeh puts different concepts of multi-valued logic into a formal system of mathematical logic, and introduced a new concept for applying natural language terms.
- This new logic (not the whole theory → Fuzzy Set Theory (FST)) for representing and manipulating fuzzy terms was called **fuzzy logic**, and Zadeh became the Master of *fuzzy logic*.

Information Fusion
Only valid with verbal explanations
Prof. Dr.-Ing. Volker Lohweg
7




Institut für Industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Introduction

- Experts rely on **common sense** when they solve problems.
- **How can we represent expert knowledge that uses vague and ambiguous terms in a computer?**
- **Fuzzy logic** (in comparison to **Boolean logic**) is not logic that is fuzzy, but logic that is used to describe fuzziness. Fuzzy logic is the theory of fuzzy sets, sets that calibrate vagueness.
- **Fuzzy Set Theory** (Fuzzy logic) is based on the idea that all things admit of degrees. Temperature, height, speed, distance, beauty – all come on a sliding scale.
 - The motor is running *really hot*.
 - Tom is a *very tall* man.

Information Fusion
Only valid with verbal explanations
Prof. Dr.-Ing. Volker Lohweg
8



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Introduction

- **Fuzzy Set Theory is a set of mathematical principles for knowledge representation based on degrees of membership.**
 - Unlike two-valued Boolean logic, fuzzy logic is **multi-valued** →


It deals with **degrees of membership** and **degrees of truth**.
 - Fuzzy logic uses the continuum of logical values between 0 (completely false) and 1 (completely true). Instead of just black and white, it employs the spectrum of colours, accepting that things can be partly true and partly false at the same time.

Information Fusion

Only valid with verbal explanations

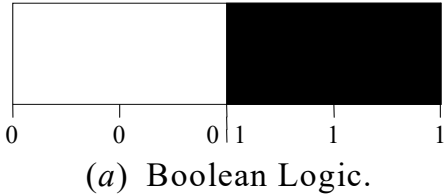
Prof. Dr.-Ing. Volker Lohweg

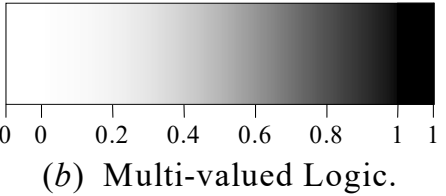
9



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Introduction





Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

10

3.3 Fuzzy Set Theory – Fuzzy Sets

- Definitions – Sets

1. Classical sets – either an element belongs to the set or it does not. For example, for the set of integers, either an integer is even or it is not (it is odd). However, either you are in Germany or you are not.

- Classical sets are also called *crisp* (sets).

Lists: $A = \{\text{apples, oranges, cherries, mangoes}\}$

$$A = \{a_1, a_2, a_3\}$$

$$A = \{2, 4, 6, 8, \dots\}$$

Formulas: $A = \{x \mid x \text{ is an even natural number}\}$

$$A = \{x \mid x = 2n, n \text{ is a natural number}\}$$

Membership or characteristic *function* $\mu_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}$

3.3 Fuzzy Set Theory – Fuzzy Sets

- Definitions – Sets

2. Fuzzy sets – admits gradation such as all tones between black and white. A fuzzy set has a graphical description which expresses how the transition from one to another takes place.

- This graphical description is called a *membership function*.

- When fuzzy variables and fuzzy sets are considered, the function describing a membership function of a variable to its set is allowed to take all values within the interval $[0 \dots 1]$.
- A fuzzy variable of an **universal set (frame of discernment)** X with a subset A is defined by its membership function $\mu_A(x)$, where $x \in R$.
- The membership function of a fuzzy variable satisfies the following properties:

$$0 \leq \mu_A(x) \leq 1$$

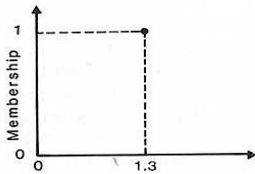
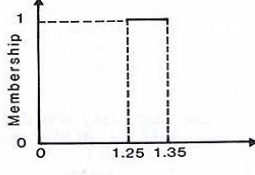
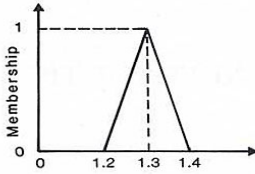
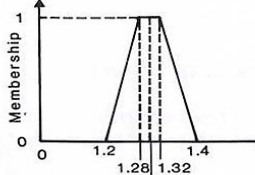
$$\mu_A(x) \text{ is convex.}$$

$$\mu_A(x) \text{ is normal. } \rightarrow \max(\mu_A(x)) = 1.$$

inIT

Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Sets

- Crisp numbers vs. Fuzzy numbers
 - (a) 
 - (b) 
 - (c) 
 - (d) 

A comparison of a real number and a crisp interval with a fuzzy number and a fuzzy interval, respectively.

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

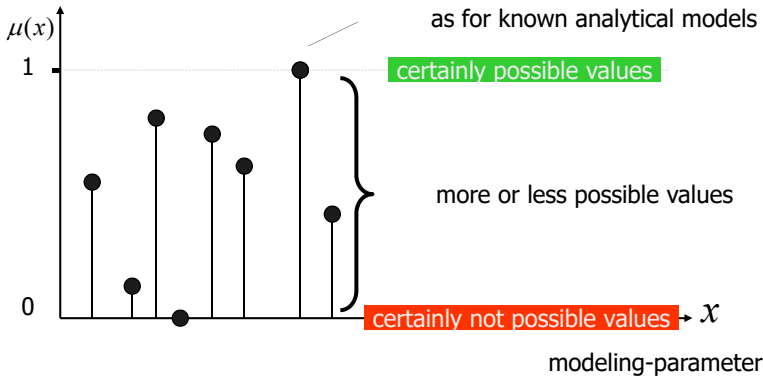
13

inIT

Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Sets

- Fuzzy Sets
 - Degree of membership to a fuzzy set



as for known analytical models

certainly possible values

more or less possible values

certainly not possible values → x

modeling-parameter

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

14

inIT

Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Sets

- Fuzzy Sets
 - Definition of discrete memberships
$$\mu_A(x) = \mu_0 / x_0 + \mu_1 / x_1 + \dots + \mu_{N-1} / x_{N-1} = \sum_{i=0}^{N-1} (\mu_i / x_i)$$
$$A = \{x_0, \dots, x_{N-1}\}$$
$$0 \leq \mu_i \leq 1$$
 - Example
$$\mu_A(x) = 0.2 / 1 + 0.6 / 2 + 1.0 / 3 + 0.4 / 4 + 0.1 / 5$$

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

15

inIT

Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

Complement

Containment

Intersection

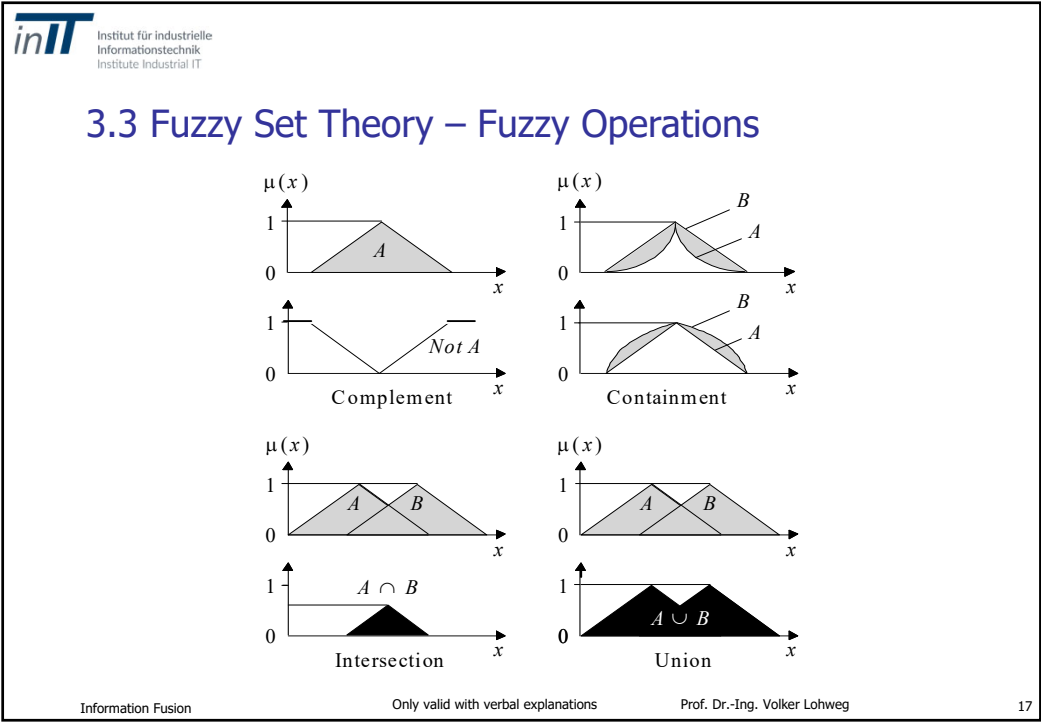
Union


Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

16





Institut für Industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

- Combination by Minimum
 - For **conjunction** (intersection) of fuzzy propositions, elements are assigned the minimum membership grade given. The fuzzy intersection


$$\mu_A(x) \cap \mu_B(x)$$

based on a t-norm, $t: [0, 1] \times [0, 1] \rightarrow [0, 1]$, is defined as:

$$\mu_{A \cap B}(x) = \min(\mu_A(x), \mu_B(x)).$$

Again, contrary to the probability theory, fuzzy union and intersection is *non-additive*.

Information Fusion
Only valid with verbal explanations
Prof. Dr.-Ing. Volker Lohweg
19



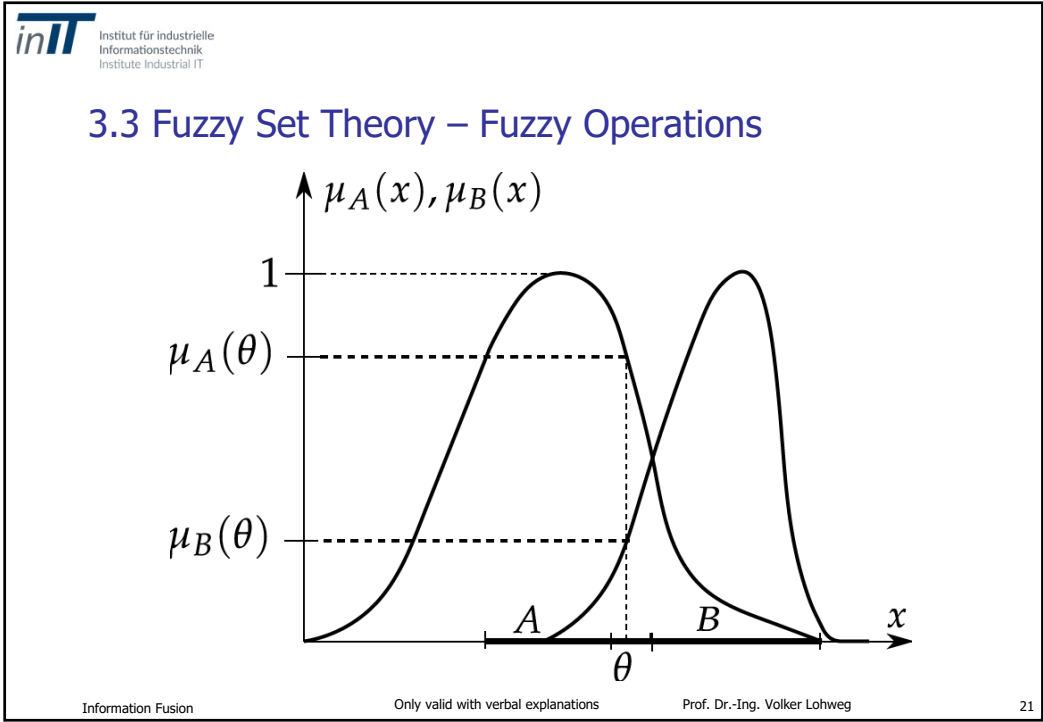
Institut für Industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

- Complement
 - The complement of a normalised membership value of the fuzzy set A is defined as:

$$\mu_{A^c}(x) = 1 - \mu_A(x).$$

Information Fusion
Only valid with verbal explanations
Prof. Dr.-Ing. Volker Lohweg
20



Institut für Industrielle Informationstechnik

Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

■ Operations – simple modifications


Nomination	Mathematical definition
Negation (NOT)	$\bar{\mu}_A(x) = \mu_{A^c}(x) = 1 - \mu_A(x)$
Disjunction (OR)	$\mu_{A \cup B}(x) = \max(\mu_A(x), \mu_B(x))$
Conjunction (AND)	$\mu_{A \cap B}(x) = \min(\mu_A(x), \mu_B(x))$
Concentration	$\mu_{kon}(x) = \mu_A^2(x)$
Reduction	$\mu_{ab}(x) = \sqrt{\mu_A(x)}$
Product	$\mu_{AB}(x) = \mu_A(x) \cdot \mu_B(x)$

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

22



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

- The membership function (MSF) of a fuzzy number or variable can also be described in terms of so called *Alpha-Cuts* (α -cuts).
- Each α -cut, at a generic level α , is defined as:


$$A_\alpha = \{ x \in A \mid \mu_A(x) \geq \alpha \}$$
- The importance of representing a fuzzy variable in terms of α -cuts is that the corresponding levels α can be considered as a **set of confidence intervals** and their associated levels of certitude.
- The **level of certitude** (certainty) contains information about how certain a human is about its knowledge.
- **“The higher the certainty the lower the confidence”** → human behaviour

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

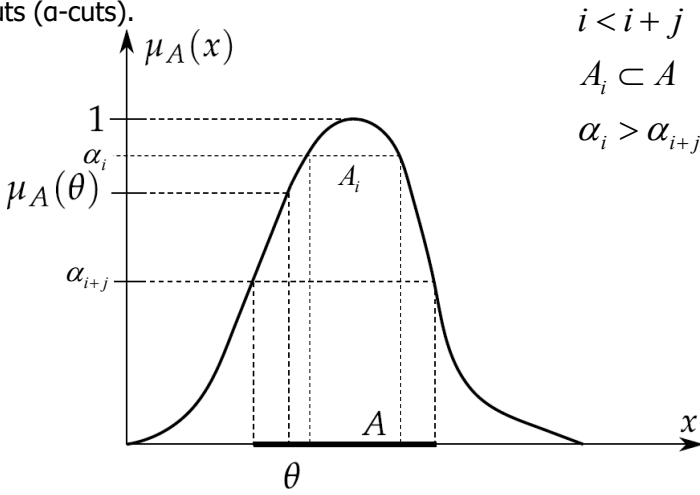
23



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

- Alpha-Cuts (α -cuts).



$i < i+j$
 $A_i \subset A$
 $\alpha_i > \alpha_{i+j}$

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

24

Information Fusion

12

3.3 Fuzzy Set Theory – Fuzzy Operations

- Fuzzy Mathematics based on cuts
 - Each fuzzy variable with finite support can fully and uniquely be represented by its α -cuts.
 - The α -cuts of each fuzzy variable are closed intervals of real numbers.
 - The operations are topic of *interval analysis*, a well-established area of classic mathematics.
- Let the symbol \circ denote any of the four arithmetic operations (addition, subtraction, multiplication and division). Then, a general property of all arithmetic operations on closed intervals is given by

$$[a, b] \circ [c, d] = \{f \circ g \mid a \leq f \leq b, c \leq g \leq d\}$$

3.3 Fuzzy Set Theory – Fuzzy Operations

- Fuzzy Mathematics based on cuts
 - Addition

$$[a, b] + [c, d] = [a + c, b + d]$$
 - Subtraction

$$[a, b] - [c, d] = [a - d, b - c]$$
 - Multiplication

$$[a, b] \cdot [c, d] = [\min(ac, ad, bc, bd), \max(ac, ad, bc, bd)]$$
 - Division

$$[a, b] / [c, d] = [a, b] \cdot \left[\frac{1}{c}, \frac{1}{d} \right] = \dots$$

$$\dots = [\min(a/c, a/d, b/c, b/d), \max(a/c, a/d, b/c, b/d)]$$

3.3 Fuzzy Set Theory – Fuzzy Operations

- Fuzzy Mathematics based on cuts – examples

- Addition

$$[2,5] + [1,3] = [3,8] \qquad [0,1] + [-6,5] = [-6,6]$$

- Subtraction

$$[0,1] - [-6,5] = [-5,7]$$

- Multiplication

$$[3,4] \cdot [2,2] = [6,8]$$


- Division

$$[4,10] / [1,2] = [2,10]$$

3.3 Fuzzy Set Theory – Fuzzy Operations

- Fuzzy Variables and Measurement


- We are able to represent measurement results with uncertainty by means of membership functions and alpha-cuts.
 - When an indirect measurement procedure is considered, the mathematics of fuzzy variables allows us to directly obtain the final result in terms of a fuzzy number (*interval and certitude*).



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

- Example 3.3-1 (cf. Example 1.2-4)
 - Again, we refer to double weighing. The variable $x_1(q)$ is the value of the first measurement which is now 2 kg, the variable $x_2(q)$ is the value of the second measurement which is now 2.4 kg. The measurement results are represented by two fuzzy variables M_1 and M_2 . Let us suppose that a contribution of not known nature, but not random, is present and the interval of confidence is +/- 100 g.




Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

29



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

- Example 3.3-1, cont'd
 - The available information is that the measurement belongs to the given interval and each point of the interval is as plausible as all others. Therefore, we have to apply rectangular membership functions and calculate the result as follows:
$$R = \frac{M_1 + M_2}{2} \rightarrow [R] = \frac{1}{2} ([1.9, 2.1] + [2.3, 2.5]) kg = \frac{1}{2} ([4.2, 4.6]) = [2.1, 2.3] kg$$
 - As already known, we have more information available. The uncertain contribution behaves systematically due to the length difference of the weighing beams. We assume the systematic error is $b \in \mathbf{R}$ in the interval of +/-100g. The double weighing will compensate the effect. It is formally:
$$R = \frac{M_1 + M_2}{2} \rightarrow [R] = \frac{1}{2} ([2 + b, 2 + b] + [2.4 - b, 2.4 - b]) kg = \frac{1}{2} ([4.4, 4.4]) = [2.2, 2.2] kg$$

CRISP!

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

30

Information Fusion

15

inIT

Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

- Fuzzy Sets – example 3.3-2 (Tall man)
 - The x -axis represents the **universe of discourse / frame of discernment** – the range of all possible values applicable to a chosen variable. In our case, the variable is the man height. According to this representation, the universe of men’s heights consists of all tall men.
 - The y -axis represents the **membership value of the fuzzy set**. In our case, the fuzzy set of “*tall men*” maps height values into corresponding membership values.

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

31

inIT

Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

Degree of Membership

Crisp Sets

Height, cm

The graph shows three crisp sets on a height axis from 150 to 210 cm. The 'Short' set is a white rectangle from 150 to 170 cm with a membership degree of 1.0. The 'Average' set is a gray rectangle from 170 to 180 cm with a membership degree of 1.0. The 'Tall' set is a black rectangle from 180 to 210 cm with a membership degree of 1.0. There are no overlaps between the sets.

Degree of Membership

Fuzzy Sets

Height, cm


The graph shows three fuzzy sets on a height axis from 150 to 210 cm. The 'Short' set is a white trapezoid starting at 150 cm (membership 1.0) and ending at 170 cm (membership 0.0). The 'Average' set is a gray triangle starting at 160 cm (membership 0.0), peaking at 175 cm (membership 1.0), and ending at 185 cm (membership 0.0). The 'Tall' set is a black trapezoid starting at 180 cm (membership 0.0) and ending at 210 cm (membership 1.0). Arrows indicate the membership values at specific heights: at 160 cm, Short is 1.0 and Average is 0.0; at 170 cm, Short is 0.0 and Average is 0.5; at 175 cm, Average is 1.0; at 180 cm, Average is 0.5 and Tall is 0.0; at 185 cm, Average is 0.0 and Tall is 0.5; at 190 cm, Tall is 1.0.

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

32

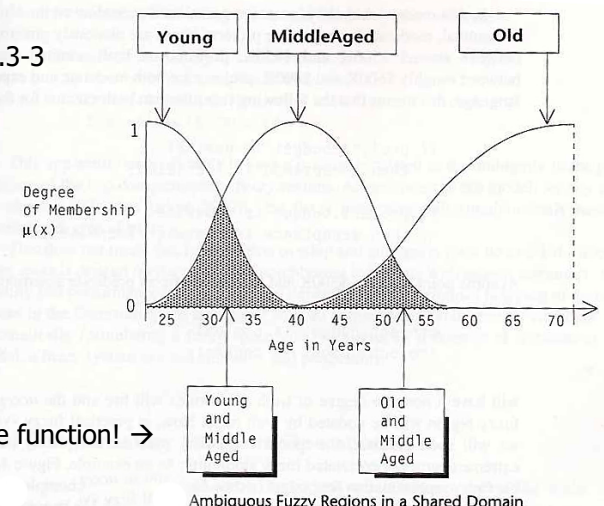


Institut für industrielle
Informationstechnik

Institute Industrial IT

3.3 Fuzzy Set Theory – Fuzzy Operations

- Fuzzy Sets – example 3.3-3



Define the function! →


Ambiguous Fuzzy Regions in a Shared Domain

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

33



Institut für industrielle
Informationstechnik

Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers


- At the root of Fuzzy Set Theory lies the idea of linguistic variables.
- A linguistic variable is a fuzzy variable.**
 - For example, the statement “John is tall” implies that the linguistic variable *John* takes the linguistic value *tall*.

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

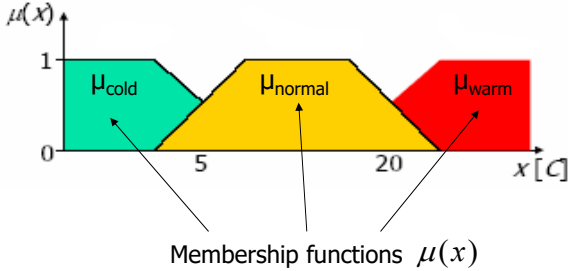
34



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers

- Linguistic Variable and Fuzzy Rules
 - A linguistic variable consists of more than one fuzzy quantities.
 - Example:
linguistic variable: temperature
linguistic term (fuzzy quantity): {cold, normal, warm}




Membership functions $\mu(x)$

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

35



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers

- Linguistic Variable and Fuzzy Rules
 - Causally determined dependencies are verbalised in form of „IF ... AND ... THEN ELSE ... IS ...“ - Rules
 - General form:
if < Premisse (assumption)> then <Conclusion>
 - linguistic variables:
 - temperature : {cold, normal, warm}
 - Price of oil : {cheap, normal, expensive}
 - Consumption of oil : {low, middle, high}
 - Fuzzy rule:
if temp is low and price of oil is cheap then consumption of oil is high


„Mathematics with words“

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

36



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers

A fuzzy rule can have multiple premisses, for example:

```
IF      < project_duration is long >
AND    < project_staffing is large >
AND    < project_funding is inadequate >
THEN   < risk is high >
```


```
IF      < service is excellent >
OR      < food is delicious >
THEN   < tip is generous >
```

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

37



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers

The conclusion of a fuzzy rule can also include multiple parts, for instance:


```
IF      < temperature is hot >
THEN   < hot_water is reduced > AND < cold_water is increased >
```

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

38



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers


- The range of possible values of a linguistic variable represents the universe of discourse of that variable. For example, the universe of discourse of the linguistic variable *speed* might have the range between 0 and 220 km/h and may include such fuzzy subsets as *very slow*, *slow*, *medium*, *fast*, and *very fast*.
- A linguistic variable carries with it the concept of fuzzy set qualifiers, called **linguistic modifiers**.
- **Linguistic modifiers** are terms that modify the shape of fuzzy sets. They include adverbs such as *very*, *somewhat*, *quite*, *more or less* and *slightly*.

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

39

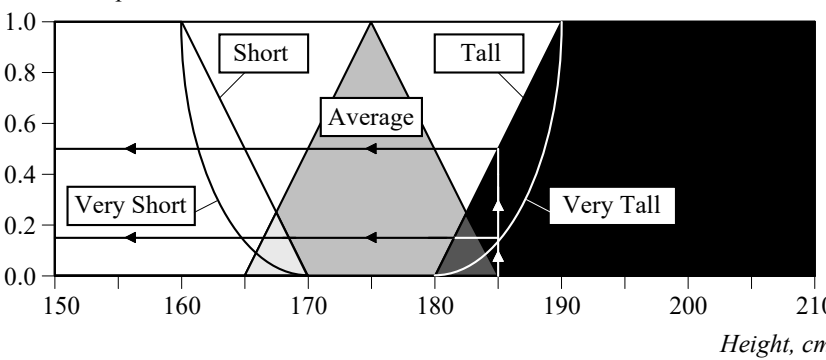


Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers

Degree of Membership

linguistic modifier: **very**



Height, cm

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

40

inTT

Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers

Adverb	Mathematical Expression	Graphical Representation
A little	$[\mu_A(x)]^{1.3}$	
Slightly	$[\mu_A(x)]^{1.7}$	
Very	$[\mu_A(x)]^2$	
Extremely	$[\mu_A(x)]^3$	

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

41

inTT

Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers

- Fuzzy Sets – example 3.3-4


Graph illustrating the Degree of Membership $\mu(x)$ versus Age in Years (25 to 55). The graph shows two fuzzy sets: Middle Age and Very Middle Age. The Very Middle Age set is a narrower, more peaked curve compared to the Middle Age set, indicating a higher degree of membership for ages closer to the center of the distribution.

Information Fusion

Only valid with verbal explanations

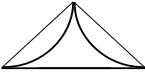
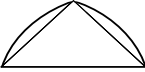
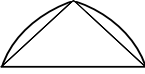

Prof. Dr.-Ing. Volker Lohweg

42



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Linguistic Variables and Modifiers


Adverb	Mathematical Expression	Graphical Representation
Very very	$[\mu_A(x)]^4$	
More or less	$\sqrt{\mu_A(x)}$	
Somewhat	$\sqrt{\mu_A(x)}$	
Indeed	$\begin{aligned} &2 [\mu_A(x)]^2 \\ &\text{if } 0 \leq \mu_A \leq 0.5 \\ &1 - 2 [1 - \mu_A(x)]^2 \\ &\text{if } 0.5 < \mu_A \leq 1 \end{aligned}$	

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

43



Institut für industrielle
Informationstechnik
Institute Industrial IT

3.3 Fuzzy Set Theory – Limitations

- Choosing Membership Functions
 - Like choosing probabilities for Bayesian approaches, in the absence of solid evidence or data, the design of membership functions can be difficult.
 - However, ... (☺) [**There is hope!**]
- This problem can be solved by using
 - **unimodal potential functions** based on
 - **measurement results.**

Information Fusion

Only valid with verbal explanations

Prof. Dr.-Ing. Volker Lohweg

44

3.3 Fuzzy Set Theory – Summary

- Fuzzy logic reflects how people think.
- It attempts to model our sense of words, our decision making and our common sense.
- As a result, it is leading to new, more human, intelligent systems.
- Like choosing probabilities for Bayesian approaches, in the absence of solid evidence or data, the design of membership functions can be difficult.