

KON 426E
INTELLIGENT CONTROL SYSTEMS

LECTURE 8
11/04/2022

Fuzzy Sets

A classical set is a set with a crisp boundary.

A classical set A of real numbers greater than 6:

$$A = \{x \mid x > 6\}$$

Now let us express the set of tall people.

Let A = "tall person" and x = "height" in cm. Define A :

$$A = \{x \mid x > 170\}$$

According to this definition a person with height 170.0001cm is tall while one with height 169.9999 is not???

So, classical (crisp) sets do not reflect the true nature of human concepts and thoughts.

- A **fuzzy set** is a set without a crisp boundary.
- The translation from “**belonging to a set**” to “**not belonging to a set**” is gradual and smooth and this is characterized by **membership functions**.
- Let X be a space of objects and x be a generic element of X .
- A (classical/ordinary/crisp/nonfuzzy) set A :
 $A \subseteq X$ is defined as a collection of elements $x \in X$
such that each x either belongs or does not belong to set A .

$$(x, 0) \quad x \notin A$$

$$(x, 1) \quad x \in A$$

- These are called **ordered pairs** and values 0 and 1 are called **characteristic functions**. In a classical set, they are either 0 or 1.

Fuzzy Sets

- Fuzzy set expresses the degree to which an element belongs to a set.

$$A = \{(x, \mu_A(x)) \mid x \in X\}$$

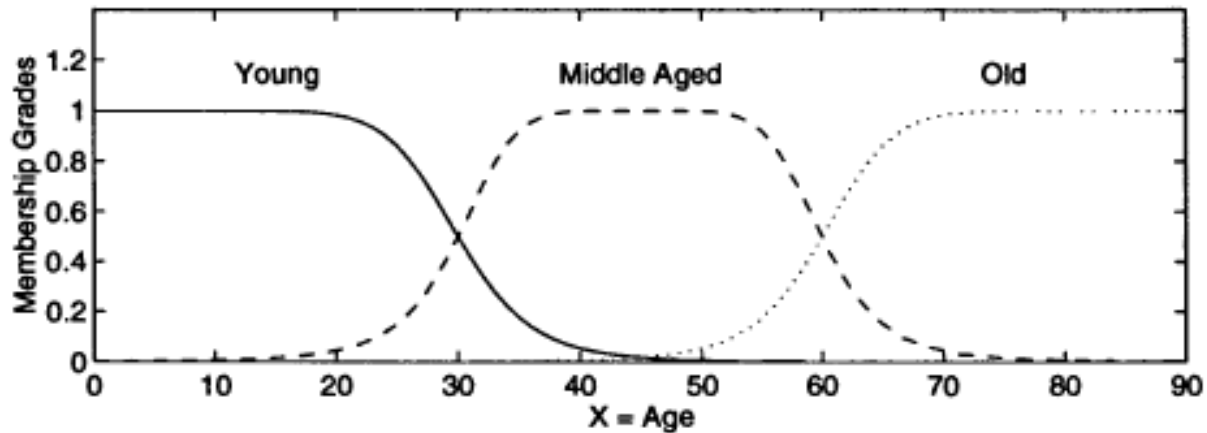
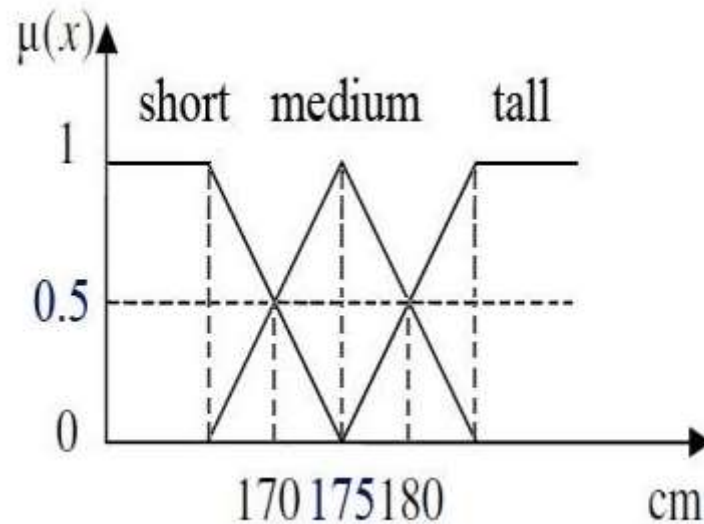
$\mu_A(x)$ is called the **membership function (MF)**

- Membership function maps each element of X to a membership value between 0 and 1.
- A classical set is a special case where the MF can take values of only 0 and 1.

X : universe of discourse

- It may consist of discrete objects or continuous space(ordered or nonordered).

Some examples of fuzzy sets and membership functions:

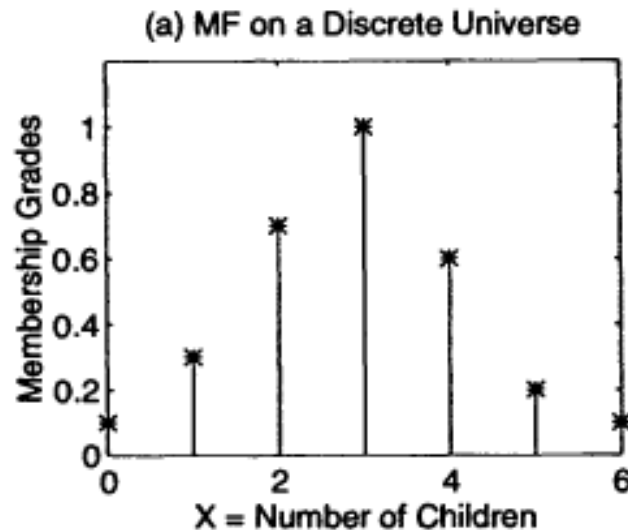


Example: Discrete ordered universe

Let $X = \{0, 1, 2, 3, 4, 5, 6\}$ be the set of numbers of children a family may choose to have. Then the fuzzy set $A =$ “sensible number of children in a family” may be described as follows:

$$A = \{(0, 0.1), (1, 0.3), (2, 0.7), (3, 1), (4, 0.7), (5, 0.3), (6, 0.1)\}.$$

Here we have a discrete ordered universe X ; the MF for the fuzzy set A is shown in Figure 2.1(a). Again, the membership grades of this fuzzy set are obviously subjective measures.



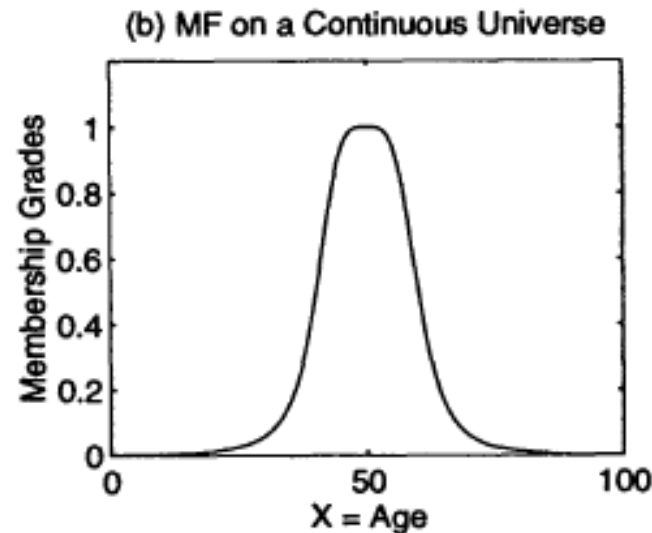
Example: Continuous ordered universe

Let $X = R^+$ be the set of possible ages for human beings. Then the fuzzy set $B =$ “about 50 years old” may be expressed as

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

where

$$\mu_B(x) = \frac{1}{1 + \left(\frac{x - 50}{10}\right)^4}.$$



Example: Discrete non-ordered universe

Let $X = \{\text{San Francisco, Boston, Los Angeles}\}$ be the set of cities one may choose to live in. The fuzzy set $C = \text{“desirable city to live in”}$ may be described as follows:

$$C = \{(\text{San Francisco}, 0.9), (\text{Boston}, 0.8), (\text{Los Angeles}, 0.6)\}.$$

Notation

$$A = \begin{cases} \sum_{x_i \in X} \mu_A(x_i) / x_i, & \text{If } X \text{ is a collection of discrete objects} \\ \int_X \mu_A(x) / x, & \text{If } X \text{ is a continuous space (e.g. the real line } R) \end{cases}$$

With this notation the previous examples can be expressed as:

$$A = 0.1/0 + 0.3/1 + 0.7/2 + 1.0/3 + 0.7/4 + 0.3/5 + 0.1/6,$$

$$B = \int_{R^+} \frac{1}{1 + \left(\frac{x-50}{10}\right)^4} / x,$$

$$C = 0.9/\text{San Francisco} + 0.8/\text{Boston} + 0.6/\text{Los Angeles},$$

Linguistic Variables and Linguistic Values

- Suppose the universe of discourse is $X = \text{"age"}$
Then **age** is a **linguistic variable**.
- We can define fuzzy sets “young”, “middle-aged”, and “old” with MF's $\mu_{old}(x)$, $\mu_{middleaged}(x)$, and $\mu_{old}(x)$
- In classical logic a variable can assume various values. e.g. $x = 10$
- In fuzzy logic, linguistic variables assume linguistic values.
e.g. “age is young” or “age is old”

Some definitions regarding fuzzy sets

Support (dayanak)

The **support** of a fuzzy set A is the set of all points x in X such that $\mu_A(x) > 0$:

$$\text{support}(A) = \{x | \mu_A(x) > 0\}$$

Core (öz)

The **core** of a fuzzy set A is the set of all points x in X such that $\mu_A(x) = 1$

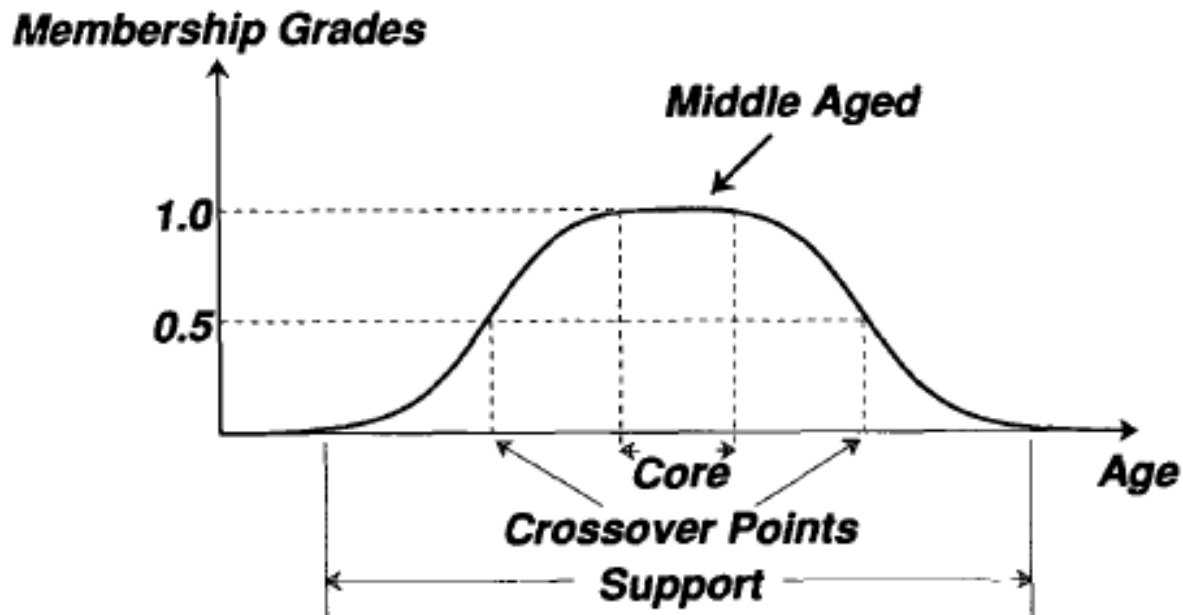
$$\text{core}(A) = \{x | \mu_A(x) = 1\}$$

$$\text{boundary}(A) = \{x | 0 \leq \mu_A(x) < 1\}$$

Normality

A fuzzy set A is **normal** if its core is nonempty.

(We can find a point $x \in X$ such that $\mu_A(x) = 1$)



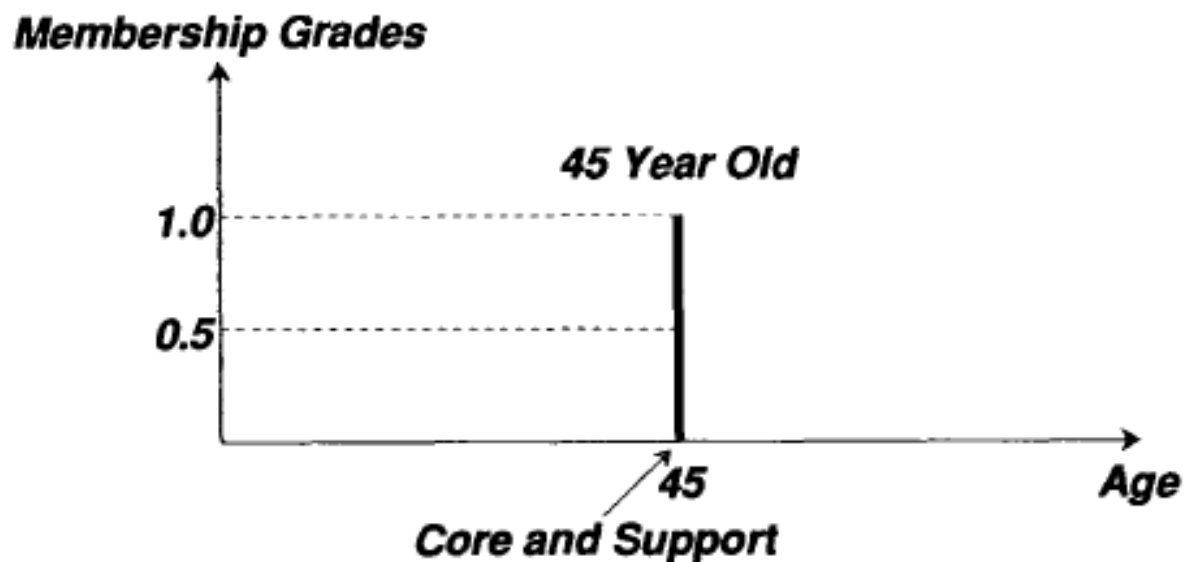
Crossover points

A **crossover point** of a fuzzy set A is a point $x \in X$ at which $\mu_A(x) = 0.5$

$$\text{crossover}(A) = \{x | \mu_A(x) = 0.5\}$$

Fuzzy Singleton

A fuzzy set whose support is a single point in X with $\mu_A(x) = 1$ is called a **fuzzy singleton**.



α -cut, strong α -cut

The α -cut (α -level set) of a fuzzy set A is a crisp set defined by:

$$A_\alpha = \{x | \mu_A(x) \geq \alpha\}$$

Strong α -cut (strong α -level set) are defined similarly:

$$A'_\alpha = \{x | \mu_A(x) > \alpha\}.$$

$$\text{support}(A) = A'_0.$$

$$\text{core}(A) = A_1$$

Convexity

- For a function $f(x)$ (where $\lambda \in [0, 1]$)

$$\lambda f(x_1) + (1 - \lambda)f(x_2) \geq f(\lambda x_1 + (1 - \lambda)x_2)$$

- For a crisp set

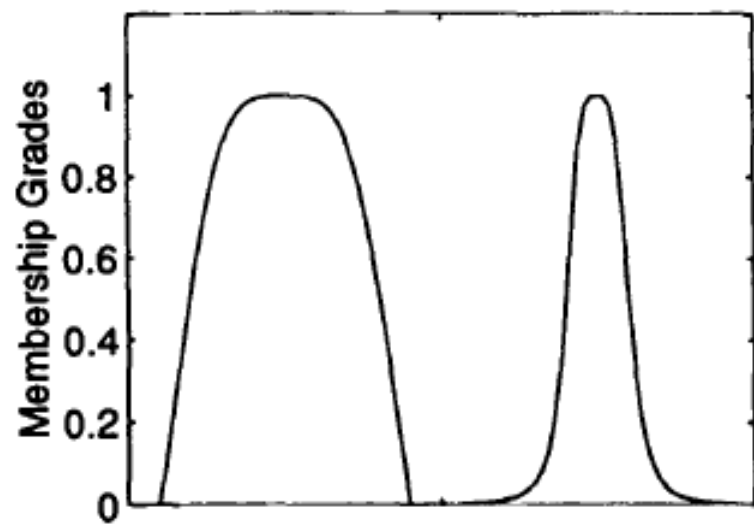
A crisp set C in R^n is convex if and only if for any two points $x_1 \in C$ and $x_2 \in C$, their convex combination $\lambda x_1 + (1 - \lambda)x_2$ is still in C , where $0 \leq \lambda \leq 1$.

- For a fuzzy set

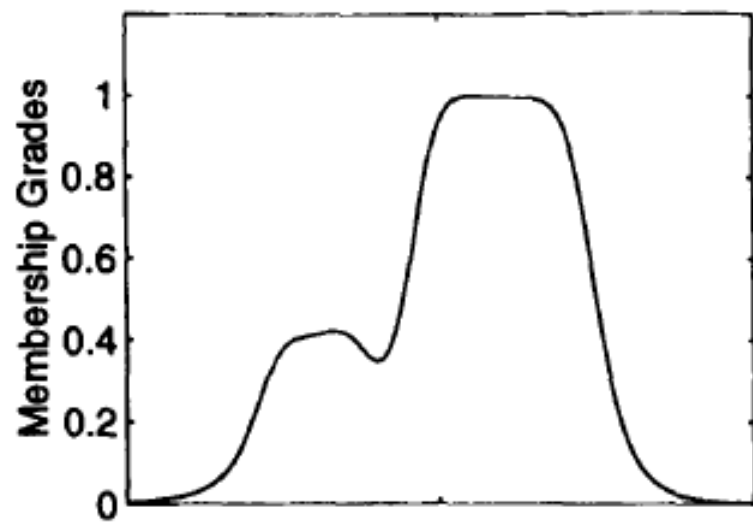
A fuzzy set A is **convex** if and only if for any $x_1, x_2 \in X$ and any $\lambda \in [0, 1]$

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

(a) Two Convex Fuzzy Sets



(b) A Nonconvex Fuzzy Set



Bandwidth for a normal and convex fuzzy set

For a normal and convex fuzzy set, the bandwidth or width is defined as the distance between two unique crossover points.

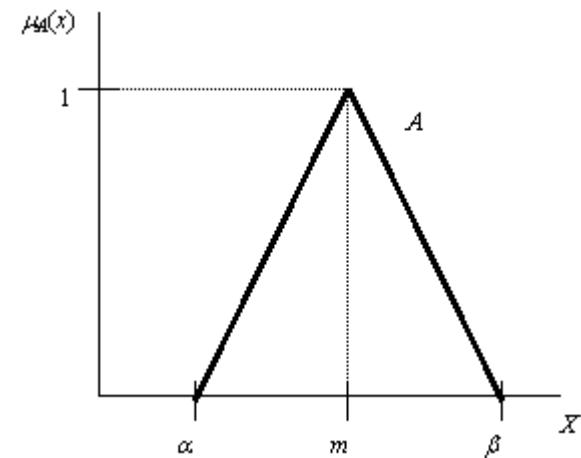
$$\text{width}(A) = |x_2 - x_1|$$

where $\mu_A(x_1) = \mu_A(x_2) = 0.5$

Symmetry

A fuzzy set A is symmetric if its MF is symmetric around a certain point $x=c$.

$$\mu_A(c + x) = \mu_A(c - x) \text{ for all } x \in X$$

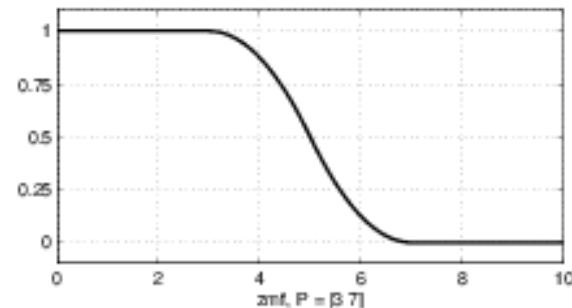


Open left, open right, closed

A fuzzy set A is **open left** if:

$$\lim_{x \rightarrow -\infty} \mu_A(x) = 1$$

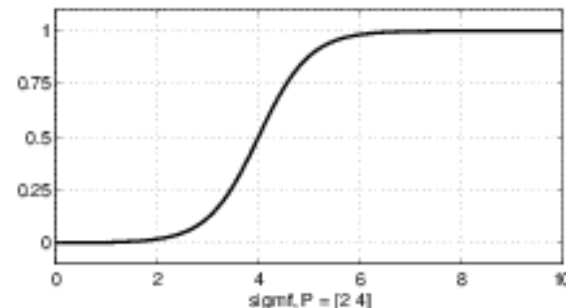
$$\lim_{x \rightarrow \infty} \mu_A(x) = 0$$



A fuzzy set A is **open right** if:

$$\lim_{x \rightarrow -\infty} \mu_A(x) = 0$$

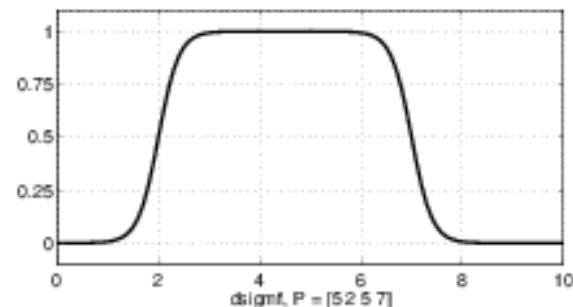
$$\lim_{x \rightarrow \infty} \mu_A(x) = 1$$



A fuzzy set A is **closed** if:

$$\lim_{x \rightarrow -\infty} \mu_A(x) = 0$$

$$\lim_{x \rightarrow \infty} \mu_A(x) = 0$$

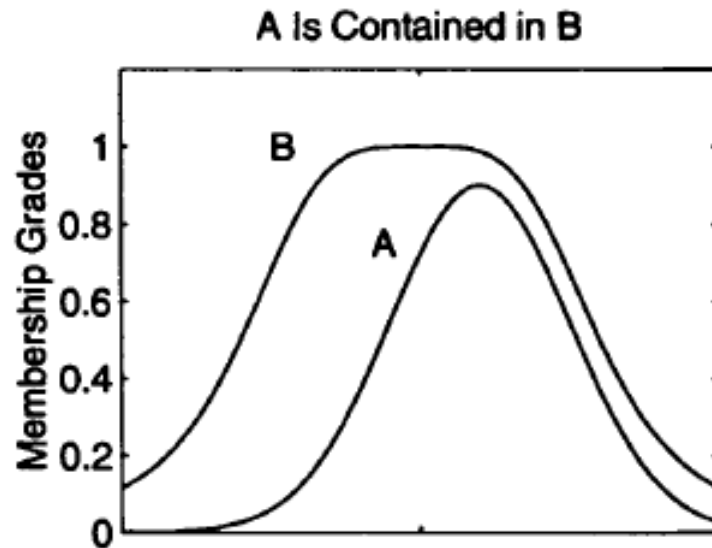


Set Theoretic Operations

Containment or subset

Fuzzy set A is contained in fuzzy set B (or A is a subset of B) if and only if $\mu_A(x) \leq \mu_B(x)$ for all x .

$$A \subseteq B \iff \mu_A(x) \leq \mu_B(x)$$



The concept of $A \subseteq B$. (MATLAB file: subset.m)

Union (Disjunction)

The union of two fuzzy sets A and B is a fuzzy set C ,

$C = A \cup B$ or $C = A \text{ OR } B$, with:

$$\mu_C(x) = \max(\mu_A(x), \mu_B(x)) = \mu_A(x) \vee \mu_B(x).$$

Union is the smallest fuzzy set containing both A and B .

Alternatively, if D is any fuzzy set that contains both A and B , then it also contains $A \cup B$

Intersection (Conjunction)

Intersection of two fuzzy sets A and B is a fuzzy set C ,

$C = A \cap B$ or $C = A \text{ AND } B$, with

$$\mu_C(x) = \min(\mu_A(x), \mu_B(x)) = \mu_A(x) \wedge \mu_B(x)$$

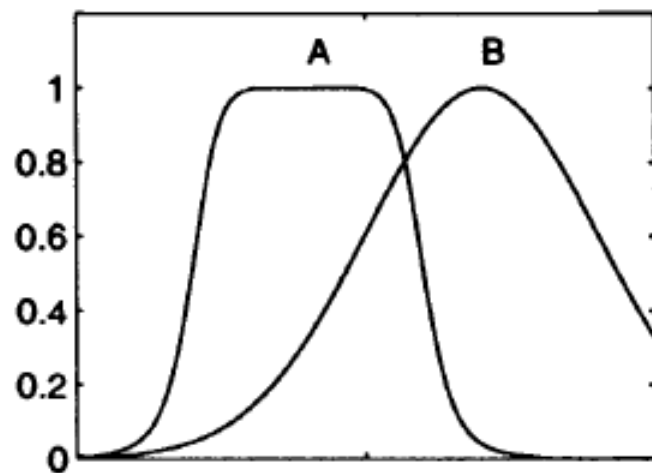
Intersection of A and B is the largest fuzzy set which is contained in both A and B .

Complement (Negation)

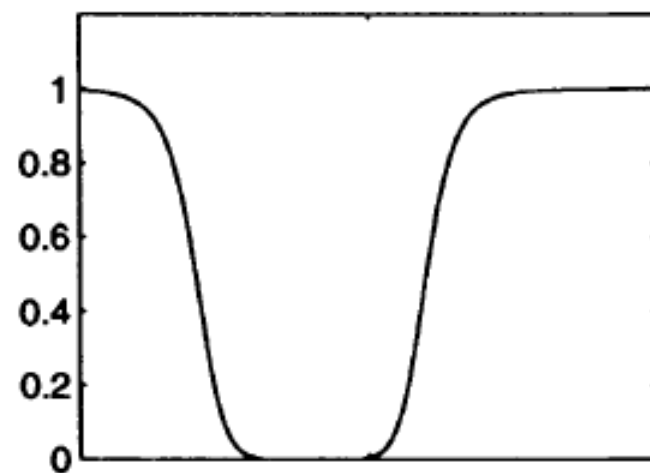
The complement of a fuzzy set A , denoted by \overline{A} ($\neg A$, NOT A) is defined as:

$$\mu_{\overline{A}}(x) = 1 - \mu_A(x)$$

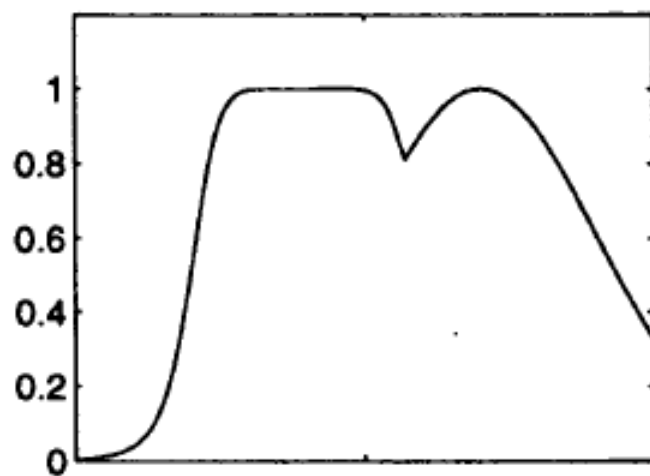
(a) Fuzzy Sets A and B



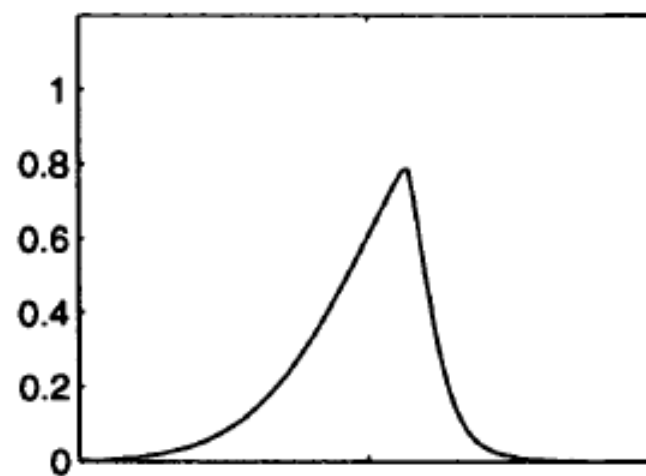
(b) Fuzzy Set "not A"



(c) Fuzzy Set "A OR B"



(d) Fuzzy Set "A AND B"



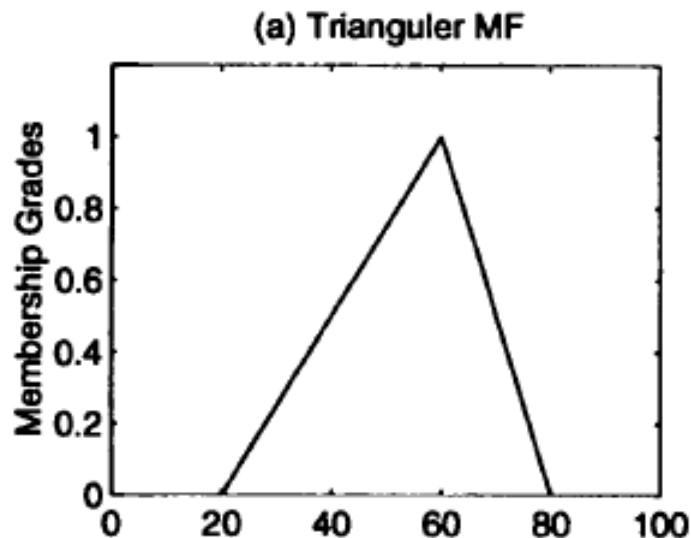
(MATLAB file: fuzsetop.m)

- Max, min and $\mu_{\overline{A}}(x) = 1 - \mu_A(x)$ operators are the classical or standard operators.
- But other operators are possible, as we will see later.
- If MF's are restricted to 0 and 1, these operators will boil down to corresponding operators for ordinary (crisp) sets.

Membership Functions of One-Dimension

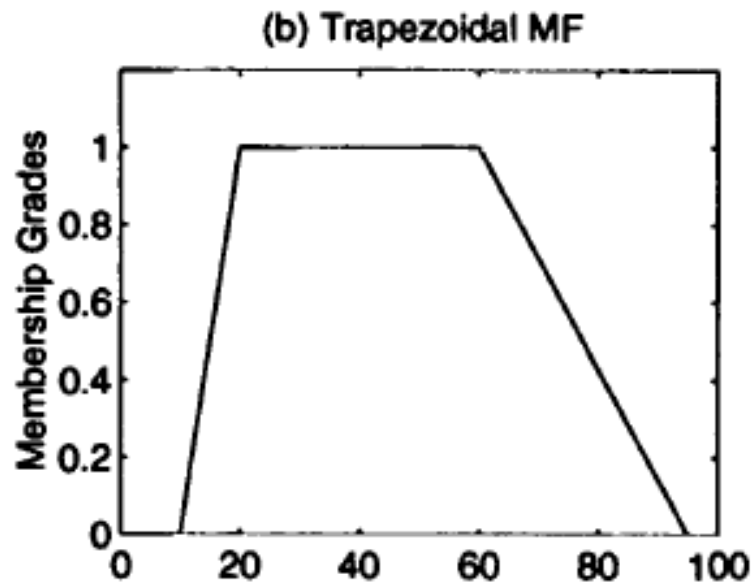
Triangular MF

$$\text{triangle}(x; a, b, c) = \begin{cases} 0, & x \leq a. \\ \frac{x-a}{b-a}, & a \leq x \leq b. \\ \frac{c-x}{c-b}, & b \leq x \leq c. \\ 0, & c \leq x. \end{cases}$$



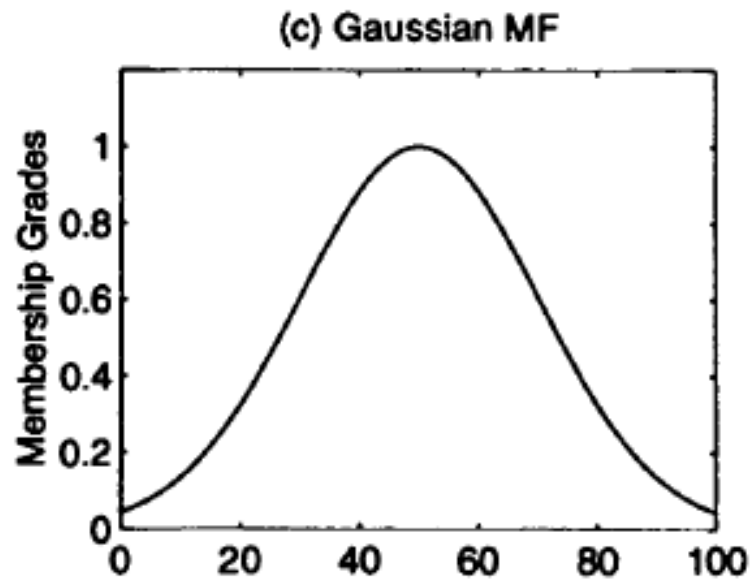
Trapezoidal MF

$$\text{trapezoid}(x; a, b, c, d) = \begin{cases} 0, & x \leq a. \\ \frac{x-a}{b-a}, & a \leq x \leq b. \\ 1, & b \leq x \leq c. \\ \frac{d-x}{d-c}, & c \leq x \leq d. \\ 0, & d \leq x. \end{cases}$$



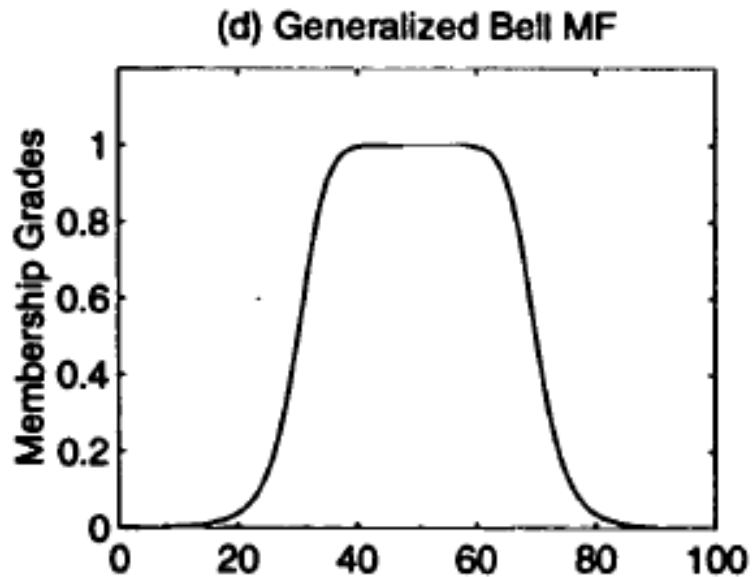
Gaussian MF

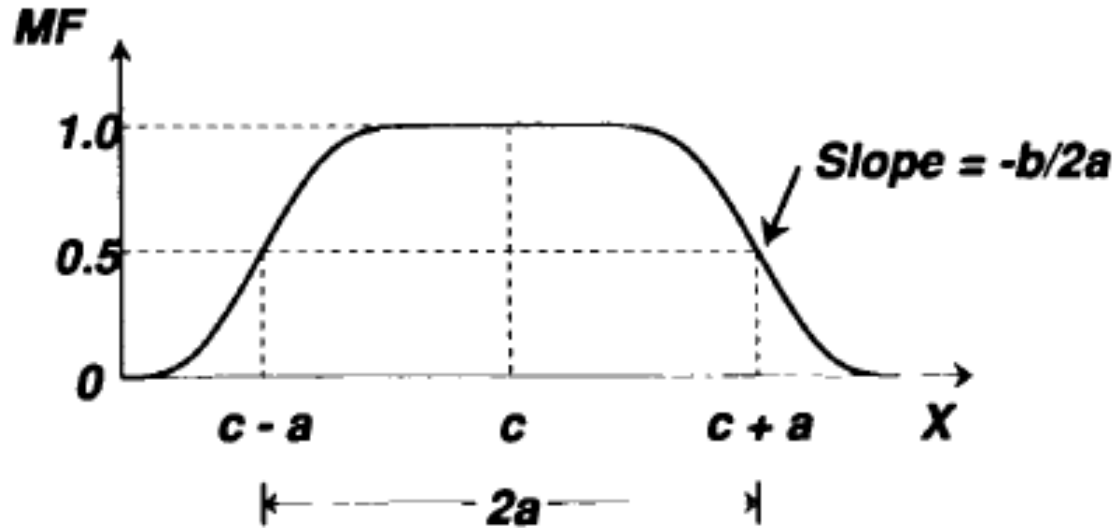
$$\text{gaussian}(x; c, \sigma) = e^{-\frac{1}{2} \left(\frac{x - c}{\sigma} \right)^2}$$



Generalized Bell MF

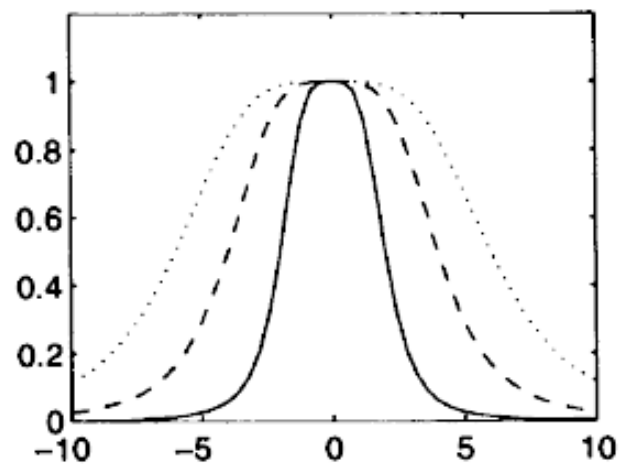
$$\text{bell}(x; a, b, c) = \frac{1}{1 + \left| \frac{x - c}{a} \right|^{2b}}$$



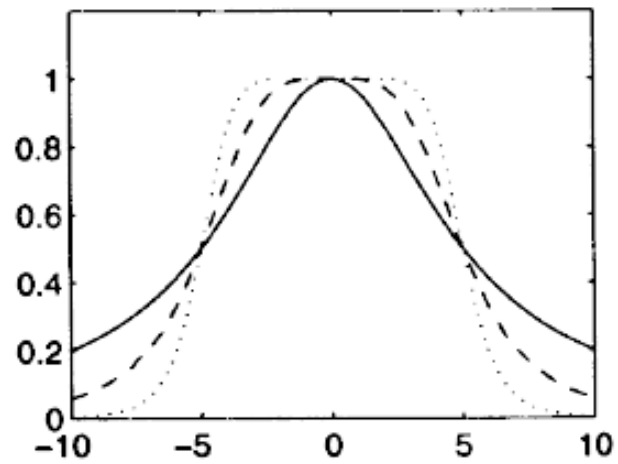


Physical meaning of parameters of the generalized bell MF

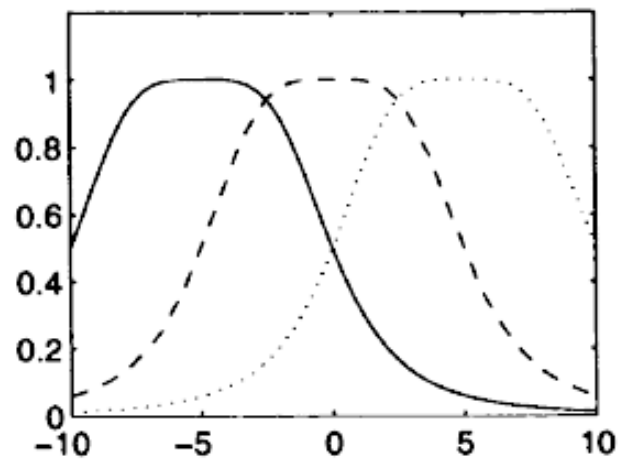
(a) Changing 'a'



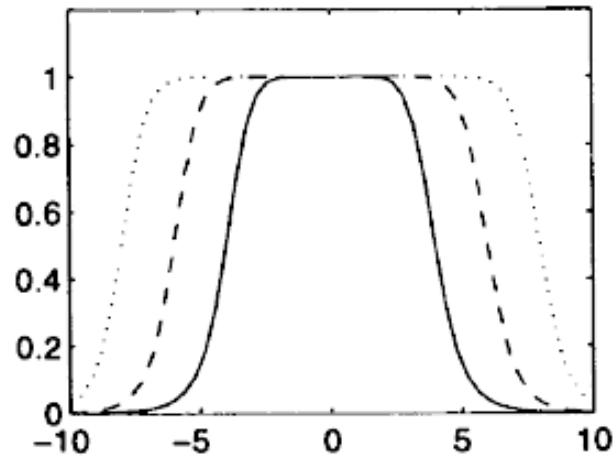
(b) Changing 'b'



(c) Changing 'c'



(d) Changing 'a' and 'b'



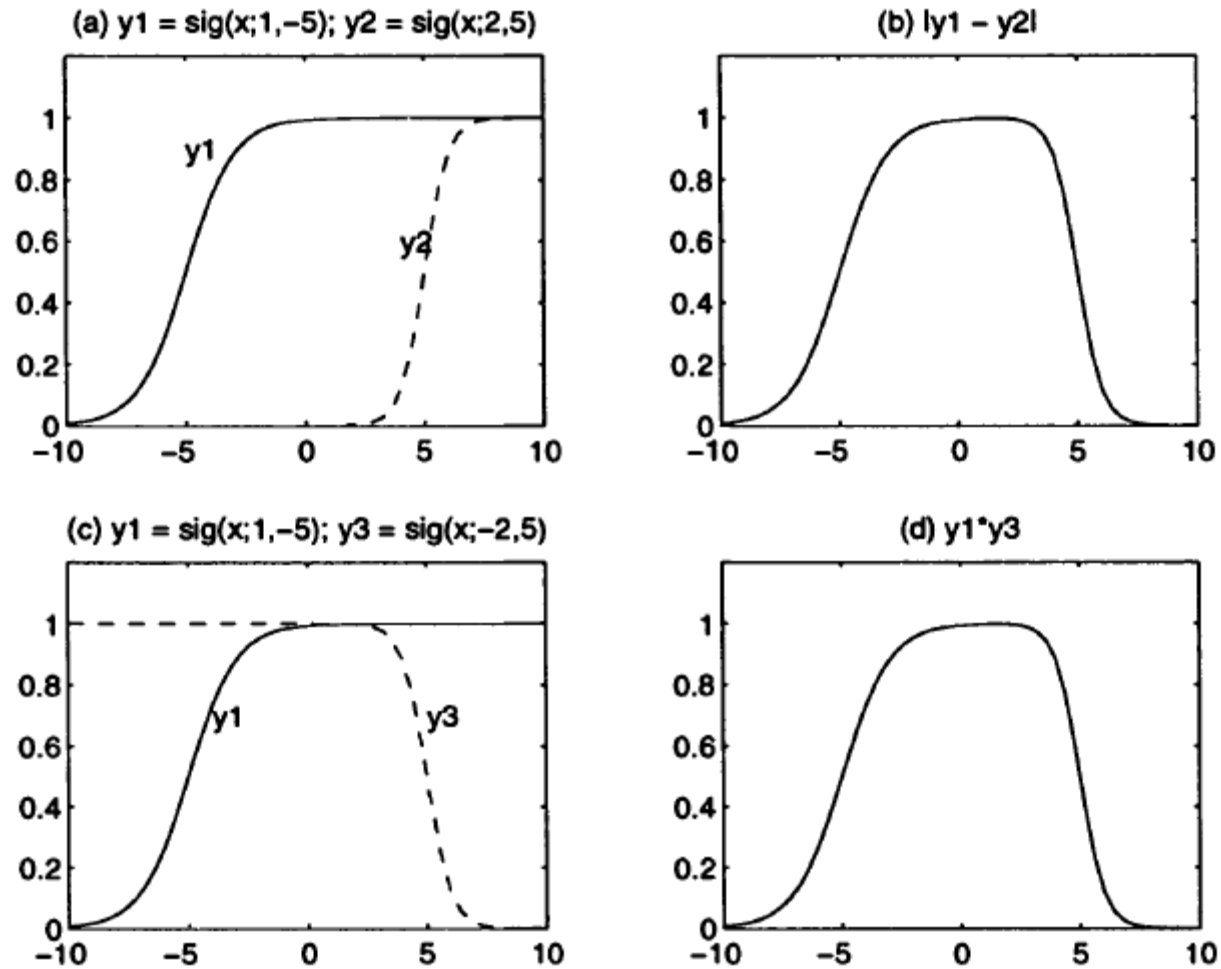
(MATLAB file: allbells.m)

Sigmoidal MF

$$\text{sig}(x; a, c) = \frac{1}{1 + \exp[-a(x - c)]}$$

where a controls the slope at the crossover point $x = c$.

Closed and asymmetric MF's based on sigmoidal functions



(MATLAB file: disp_sig.m)

Parameterized Membership Functions

S-shaped MF

smf

S-shaped membership function

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[collapse all in page](#)

Syntax

```
y = smf(x,params)
```

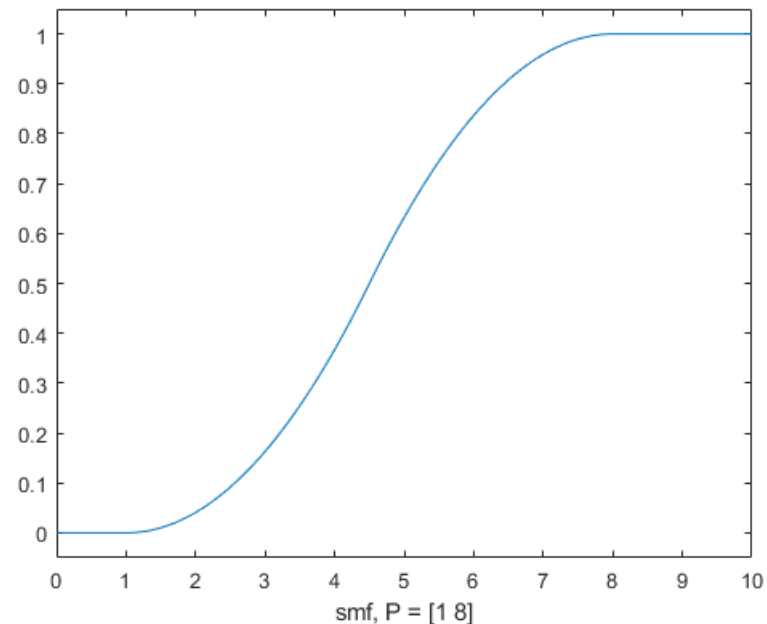
Description

This function computes fuzzy membership values using a spline-based S-shaped membership function. You can also compute this membership function using a `fismf` object. For more information, see [fismf Object](#).

This membership function is related to the [zmf](#) and [pimf](#) membership functions

`y = smf(x,params)` returns fuzzy membership values computed using the spline

```
x = 0:0.1:10;  
y = smf(x,[1 8]);  
plot(x,y)  
xlabel('smf, P = [1 8]')  
ylim([-0.05 1.05])
```



[example](#)

Z-shaped MF

zmf

Z-shaped membership function

Syntax

```
y = zmf(x,params)
```

Description

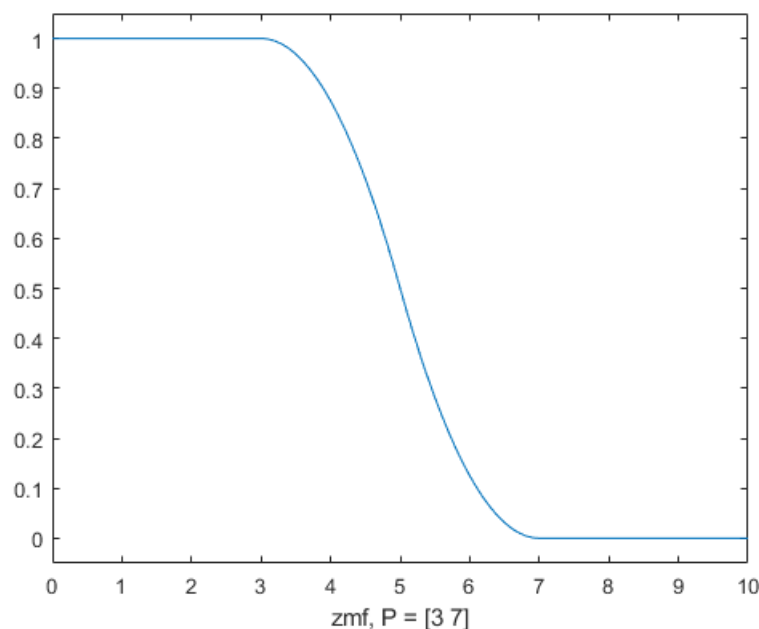
This function computes fuzzy membership values using a spline-based Z-shaped membership function. You can also compute this membership function using a `fismf` more information, see [fismf Object](#).

This membership function is related to the [smf](#) and [pimf](#) membership functions.

`y = zmf(x,params)` returns fuzzy membership values computed using the spline-based Z-shaped membership function given by:

$$f(x; a, b) = \begin{cases} 1, & x \leq a \\ 1 - 2\left(\frac{x-a}{b-a}\right)^2, & a \leq x \leq \frac{a+b}{2} \\ 2\left(\frac{x-b}{b-a}\right)^2, & \frac{a+b}{2} \leq x \leq b \\ 0, & x \geq b \end{cases}$$

To specify the a and b parameters, use `params`.



Pi-shaped MFs

Pi-shaped membership function

Syntax

```
y = pimf(x,params)
```

Description

This function computes fuzzy membership values using a spline-based pi-shaped membership function. You can also compute this membership function using more information, see [fismf Object](#).

This membership function is related to the [smf](#) and [zmf](#) membership functions.

`y = pimf(x,params)` returns fuzzy membership values computed using a spline-based pi-shaped membership function. This membership function is the pro [smf](#) function and a [zmf](#) function, and is given by:

$$f(x;a,b,c,d) = \left\{ \begin{array}{ll} 0, & x \leq a \\ 2\left(\frac{x-a}{b-a}\right)^2, & a \leq x \leq \frac{a+b}{2} \\ 1 - 2\left(\frac{x-b}{b-a}\right)^2, & \frac{a+b}{2} \leq x \leq b \\ 1, & b \leq x \leq c \\ 1 - 2\left(\frac{x-c}{d-c}\right)^2, & c \leq x \leq \frac{c+d}{2} \\ 2\left(\frac{x-d}{d-c}\right)^2, & \frac{c+d}{2} \leq x \leq d \\ 0, & x \geq d \end{array} \right\}$$

