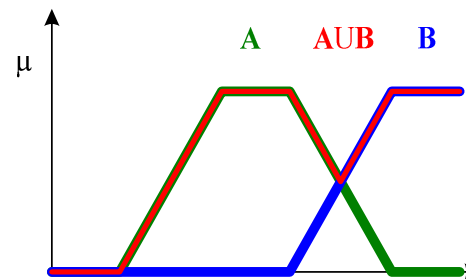
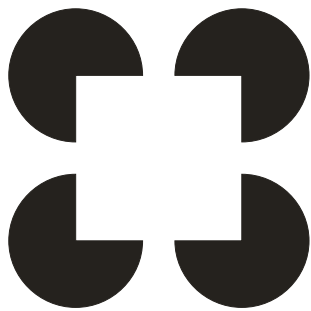
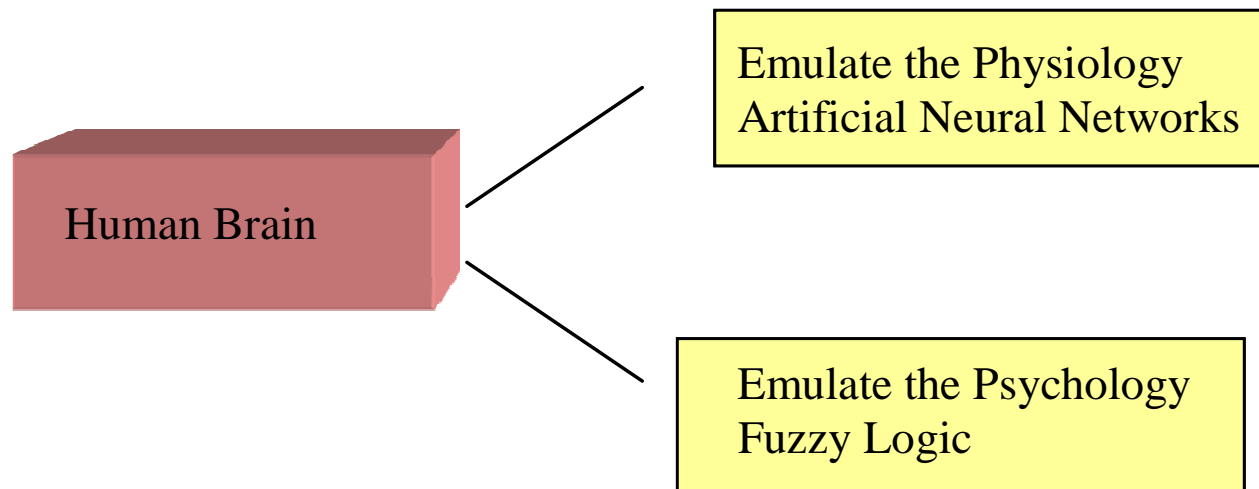


Part 3 –Fuzzy Logic and Fuzzy Systems



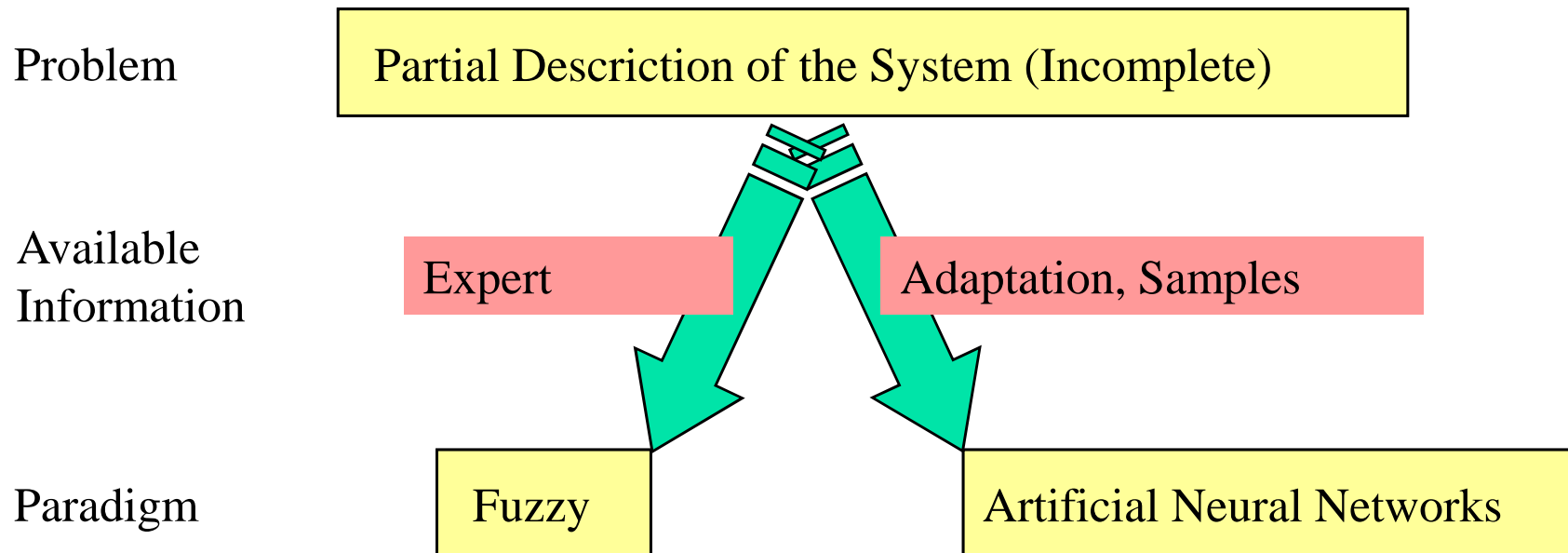
Fuzzy Logic

- The fuzzy set theory was proposed by Lotfi Zadeh in 1965.
- Was long misunderstood.
- In the mid-80s used to design Mamdani fuzzy controllers

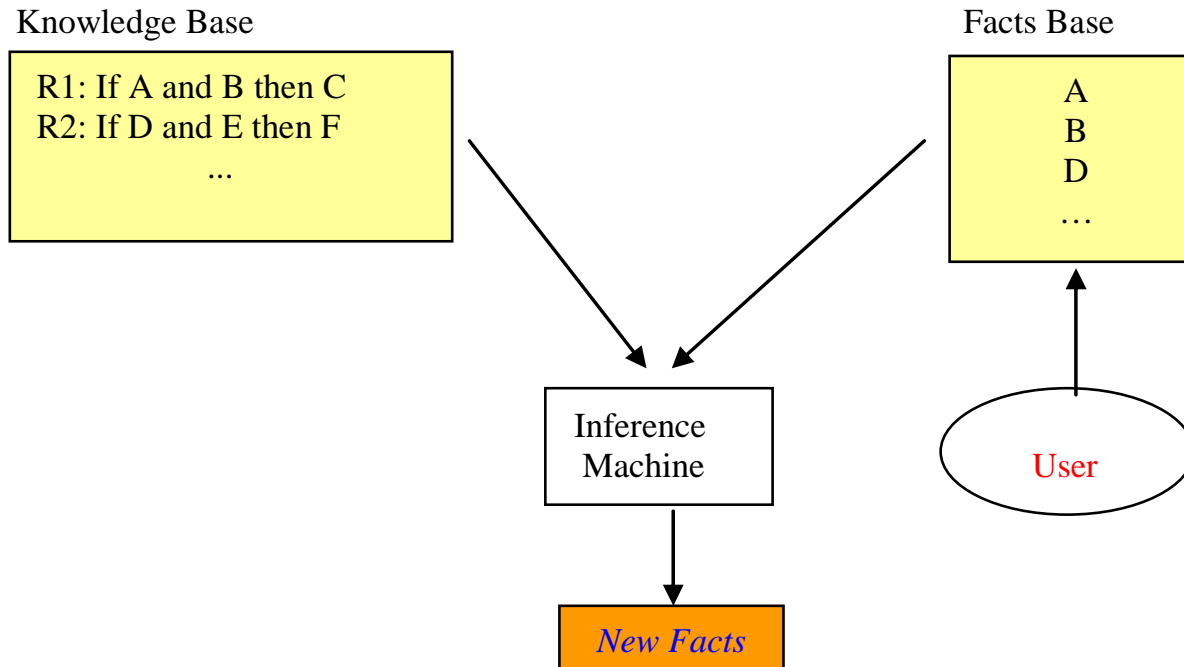


Fuzzy Logic

According to the availability of an *expert* Or *samples* of a system the fuzzy or the RNA paradigm is indicated.



Expert Systems

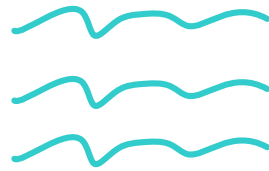


1. Extraction of Knowledge of Expert (build knowledge base)
2. I.T. expert creates the environment (shell)
3. “Normal” human presents facts and questions (over and over)
4. Response of ES similar to the human expert!

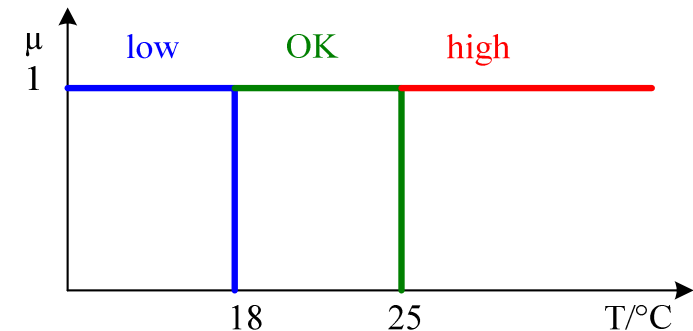
Example of “If-Then” Rules



Air Conditioner

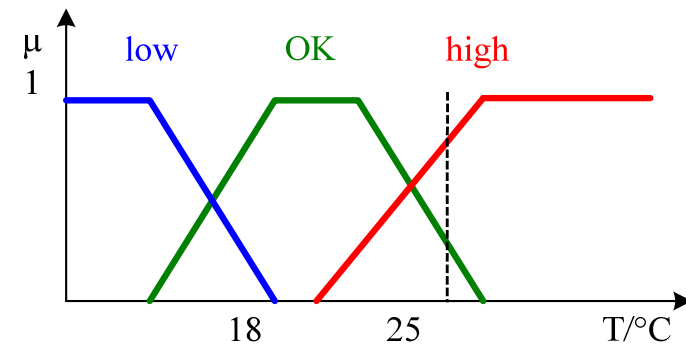


Conventional Sets



- If** temperature is **low**
Then reduce air conditioning
- If** temperature is **OK**
Then do nothing
- If** temperature is **high**
Then increase air conditioning power

Fuzzy Sets



Partial membership to both linguistic concepts!

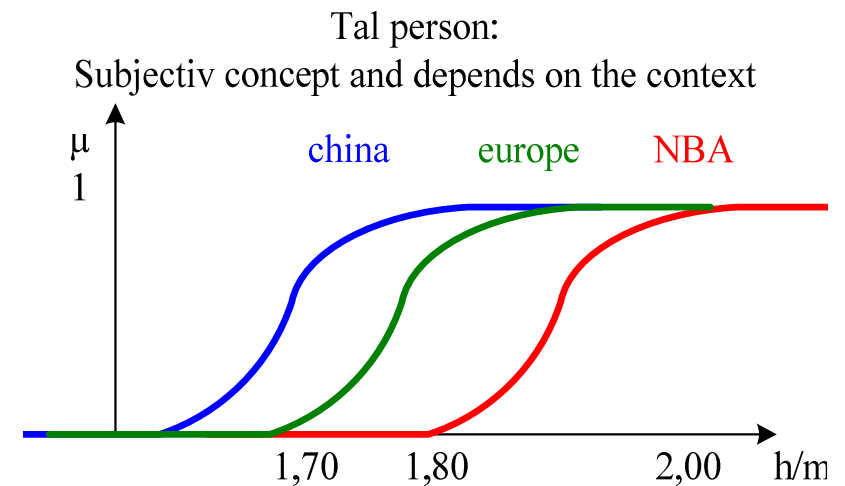
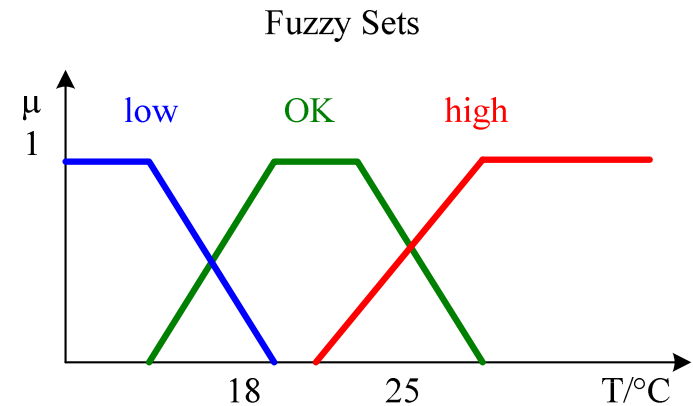
Fuzzy Sets – Membership Function

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

$$\mu_A(x) = \begin{cases} 1 & \text{if } x \text{ is a fully member of } A \\ (0,1) & \text{if } x \text{ belongs partially to } A \\ 0 & \text{if } x \text{ is not a member of } A \end{cases}$$

Extension of the Boolean Logic – Historical perspective:

- ~ 1930, Lukasiewicz : $\{0,1/2,1\}$, $[0,1]$
- 1937, Black : Membership function
- 1965, Lotfi Zadeh : Fuzzy Sets
Multivalent Set Theory
- ~ 1988, Commercial Products : “third wave” of interest



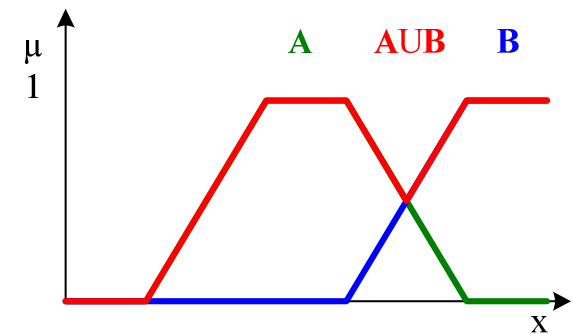
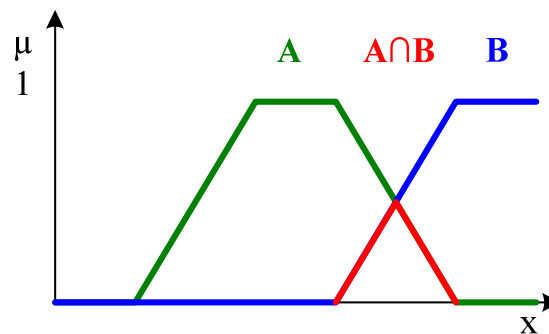
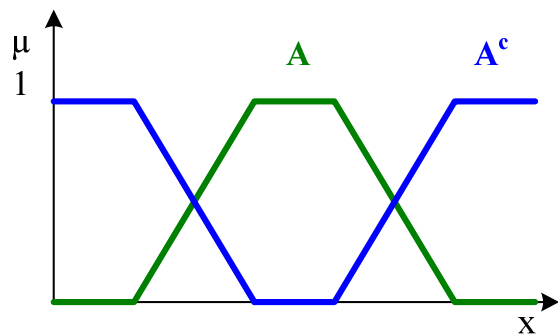
Fuzzy Sets – operations

Ex.: Complement, Intersection and Union

$$\mu A^c(x) = 1 - \mu A(x)$$

$$\mu A \cap B(x) = \min(\mu A(x), \mu B(x))$$

$$\mu A \cup B(x) = \max(\mu A(x), \mu B(x))$$



Fuzzy Sets – Properties

Involution	$(A^c)^c = A$	
Comutativity	$A \cup B = B \cup A$	$A \cap B = B \cap A$
Associativity	$A \cup (B \cup C) = (A \cup B) \cup C$	$A \cap (B \cap C) = (A \cap B) \cap C$
Distributivity	$A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$	$A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$
Idempotency	$A \cup A = A$	$A \cap A = A$
Absorption	$A \cup (A \cap B) = A$	$A \cap (A \cup B) = A$
Identity	$A \cup \Phi = A$	$A \cap \Omega = A$
Absorption by Ω e Φ	$A \cup \Omega = \Omega$	$A \cap \Phi = \Phi$
De Morgan's Law	$(A \cap B)^c = A^c \cup B^c$	$(A \cup B)^c = A^c \cap B^c$

However:

$$A \cap A^c \neq \Phi$$

Does not satisfy the law of no-contradiction

$$A \cup A^c \neq \Omega$$

Does not satisfy the law of third excluded

$$A \cup (A^c \cap B) \neq A \cup B$$

Does not satisfy the absorption of the complement

$$A \cap (A^c \cup B) \neq A \cap B$$

Does not satisfy the absorption of the complement

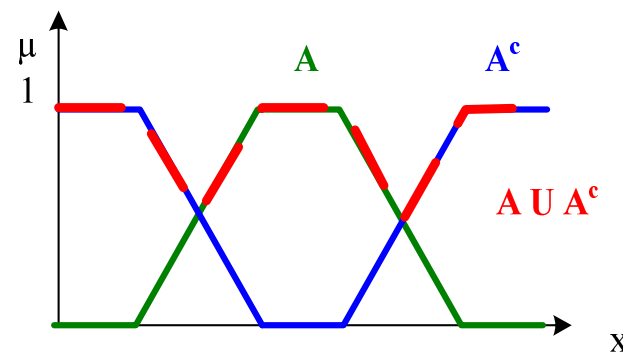
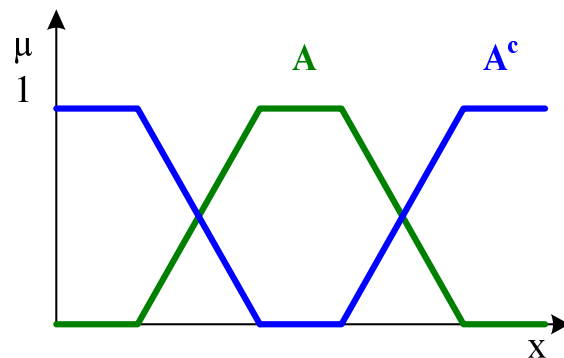
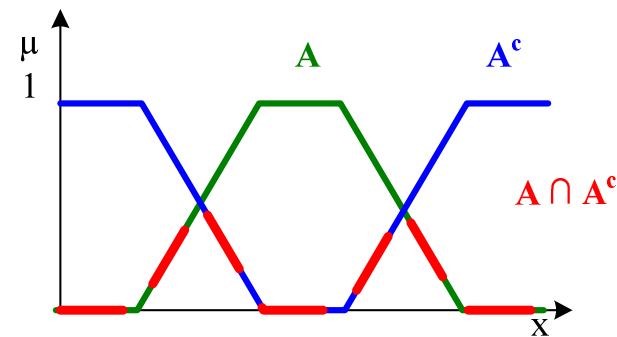
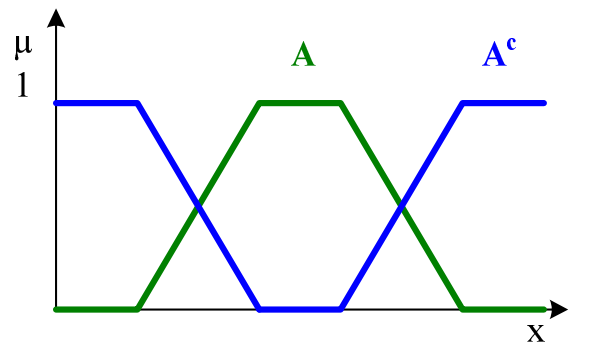
Fuzzy Sets – Examples of “strange” behavior

$$A \cap A^c \neq \Phi$$

$$A \cup A^c \neq \Omega$$

Does not satisfy the law of non-contradiction

Does not satisfy of the third excluded



Sentence Calculus – Classical Logic

In Classical Logic, the truth values of propositions (sentence calculus) are obtained by the following truth table ("modus ponens" – affirmative modus).

A	B	$\neg A$	$A \wedge B$	$A \vee B$	$A \rightarrow B$ ($\neg A \vee B$)
0	0	1	0	0	1
0	1	1	0	1	1
1	0	0	0	1	0
1	1	0	1	1	1

Sentence Calculus – Fuzzy Logic

When the information is inaccurate, the inference engine implements the so-called approximate reasoning.

Fuzzy logic implements approximate reasoning in the context of fuzzy sets ("generalized modus ponens").

fact:	A'	Tomatoes are very red
rule:	$A \rightarrow B$	If tomatoes are red then they are mature
<hr/>		
consequence	B'	The tomatoes are very ripe

$\neg A$	=	$n(A)$	n – negation
$A \wedge B$	=	$T(A,B)$	T – t -norm
$A \vee B$	=	$S(A,B)$	S – t-conorm
$A \rightarrow B$	=	$I(A,B)$	I – implication

Implication Operators

“If <premise> Then <conclusion>”

$I : [0,1]^2 \rightarrow [0,1]$, $\mu_A : X \rightarrow [0,1]$, $\mu_B : Y \rightarrow [0,1]$

$\mu_{A \rightarrow B}(x,y) = I(\mu_A(x), \mu_B(y))$

Implication	Name
$\max(1-a,b)$	Kleene-Dimes
$\min(1-a+b,1)$	Lukasiewicz
$\min(a,b)$	Mamdani
$a \cdot b$	Larsen
...	



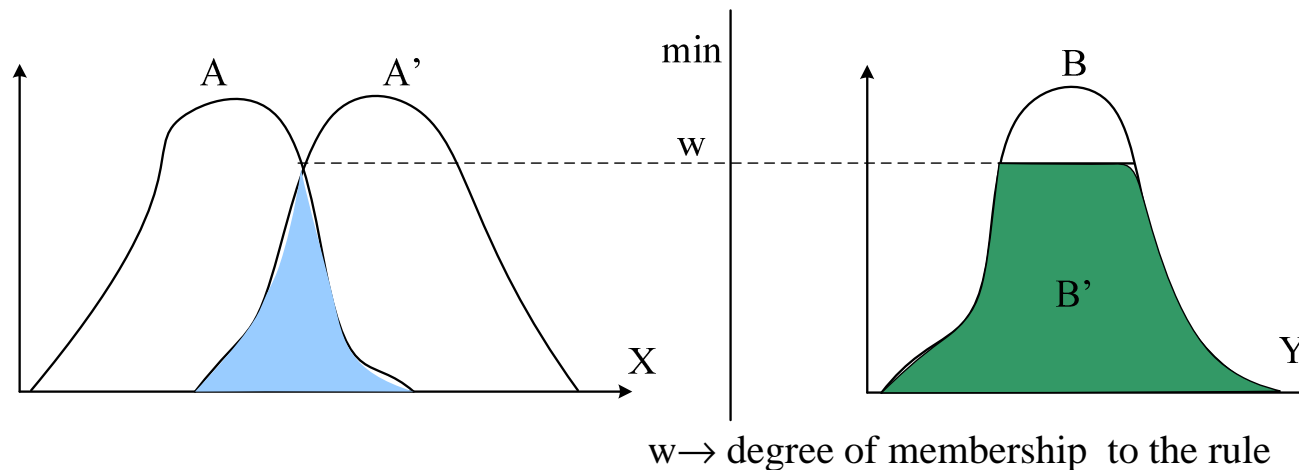
Fuzzy Reasoning

based on Max-Min composition

Definition: Let A, A' and B fuzzy sets on X, X and Y respectively. Assume that the fuzzy implication $A \rightarrow B$ is expressed by the fuzzy relation R on $X \times Y$, then the fuzzy set B' is induced "x is A' " and the fuzzy rule "if x is A then y is B " is defined by:

$$\begin{aligned} \mu_{B'}(y) &= \max_x \min [\mu_{A'}(x), \mu_R(x,y)] \\ &= \bigvee_x [\mu_{A'}(x) \wedge \mu_R(x,y)], \quad \text{that means: } B' = A' \circ R = A' \circ (A \rightarrow B) \end{aligned}$$

One fuzzy rule with one antecedent

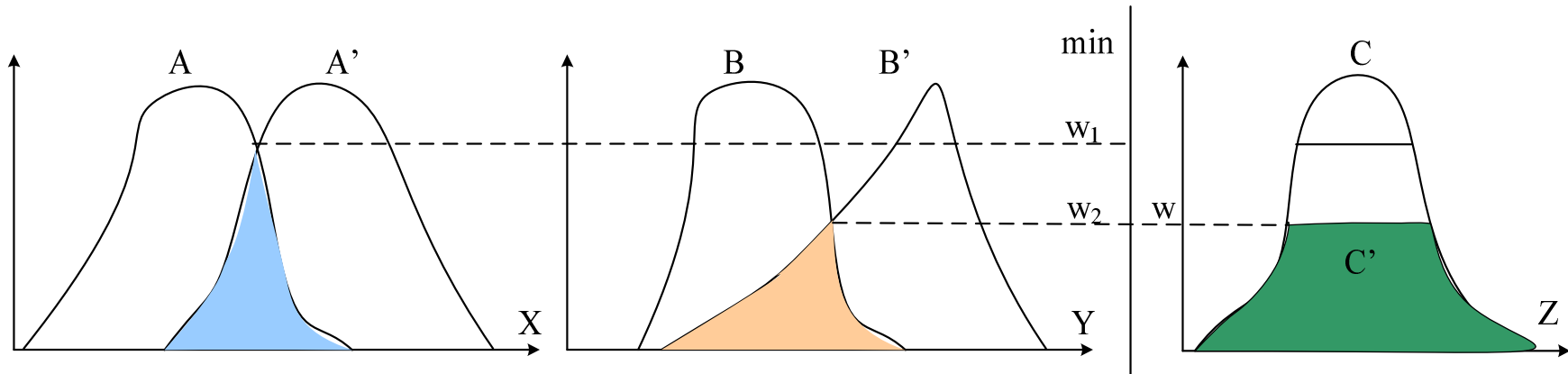


Fuzzy Reasoning

based on Max-Min composition

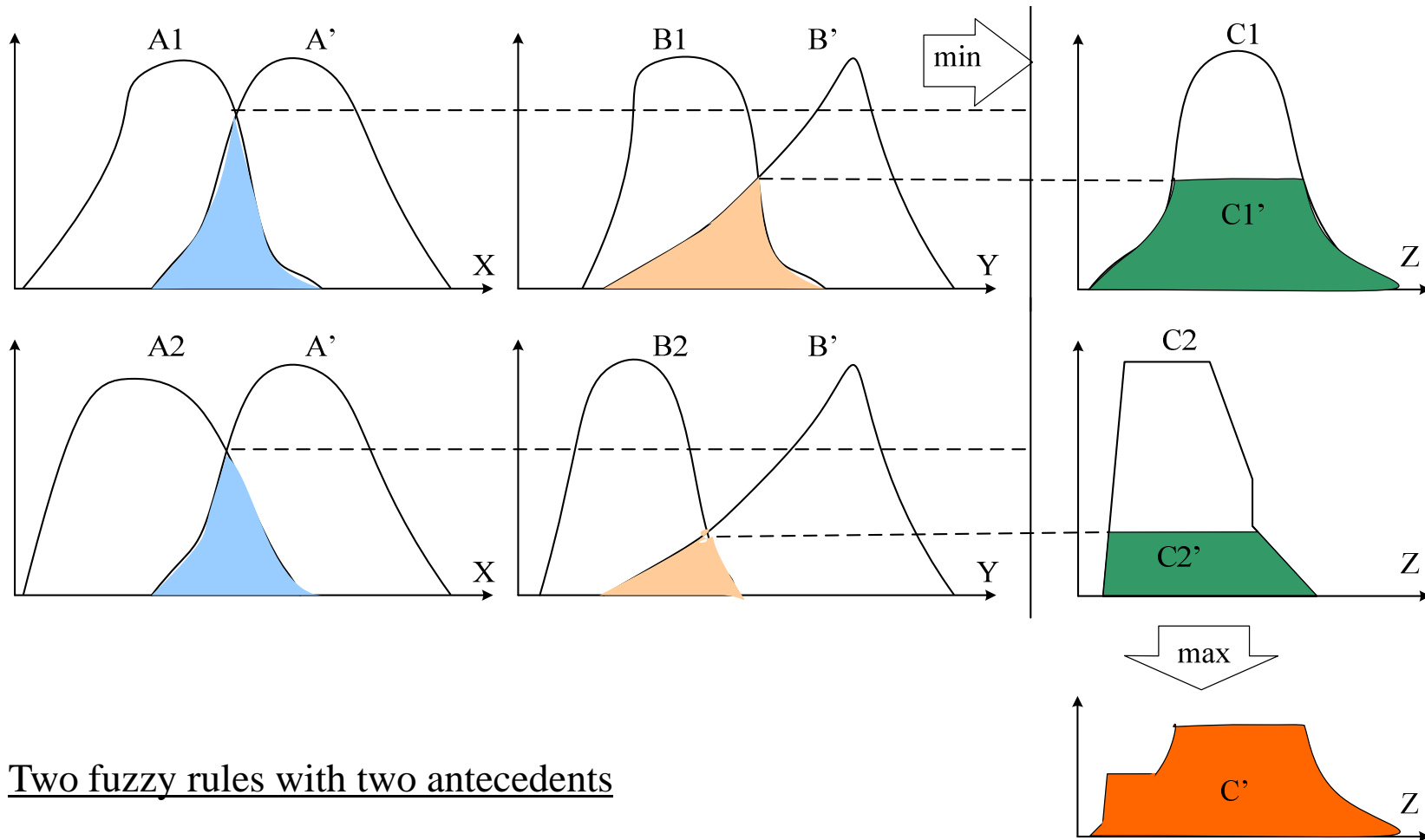
One fuzzy rule with two antecedents

“if x is A and y is B then z is C”



$w_1, w_2 \rightarrow$ degrees of membership to the respective rules.

Max-Min Fuzzy Reasoning



Two fuzzy rules with two antecedents

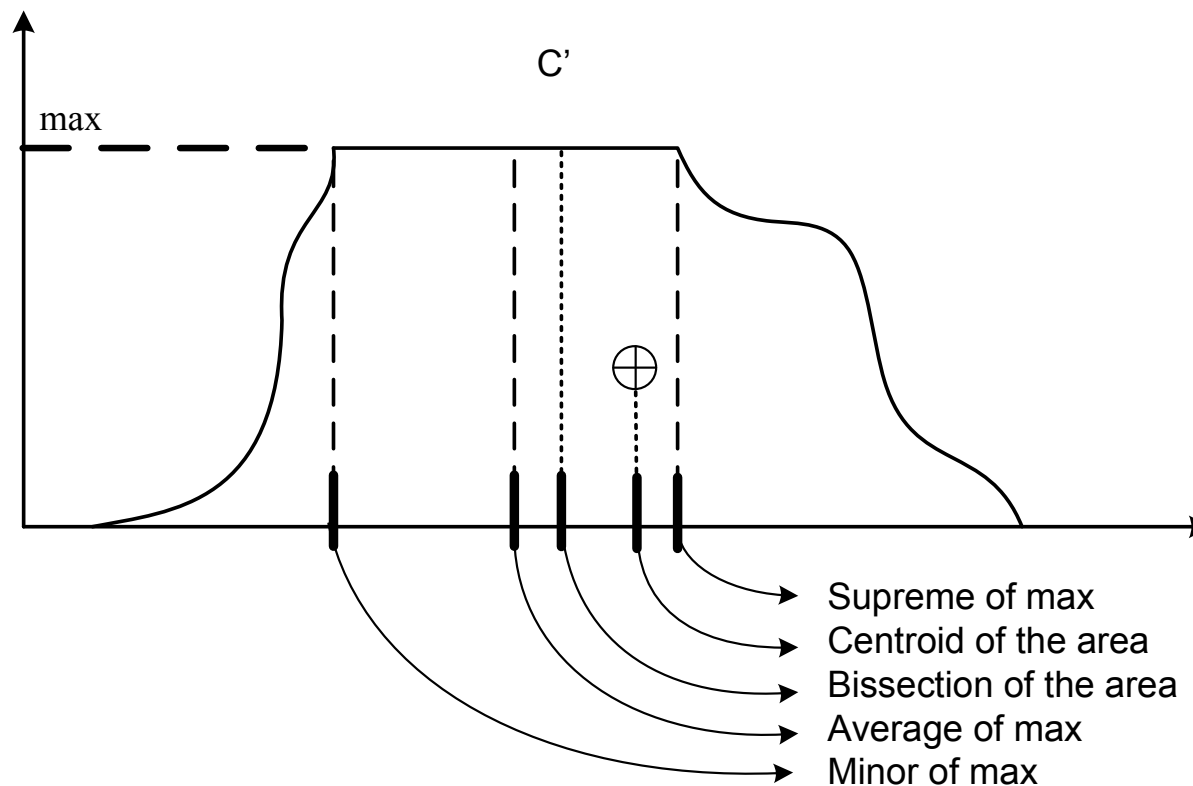
“If x is A_1 and y is B_1 then z is C_1 ”

“If x is A_2 and y is B_2 then z is C_2 ”

Result: C'

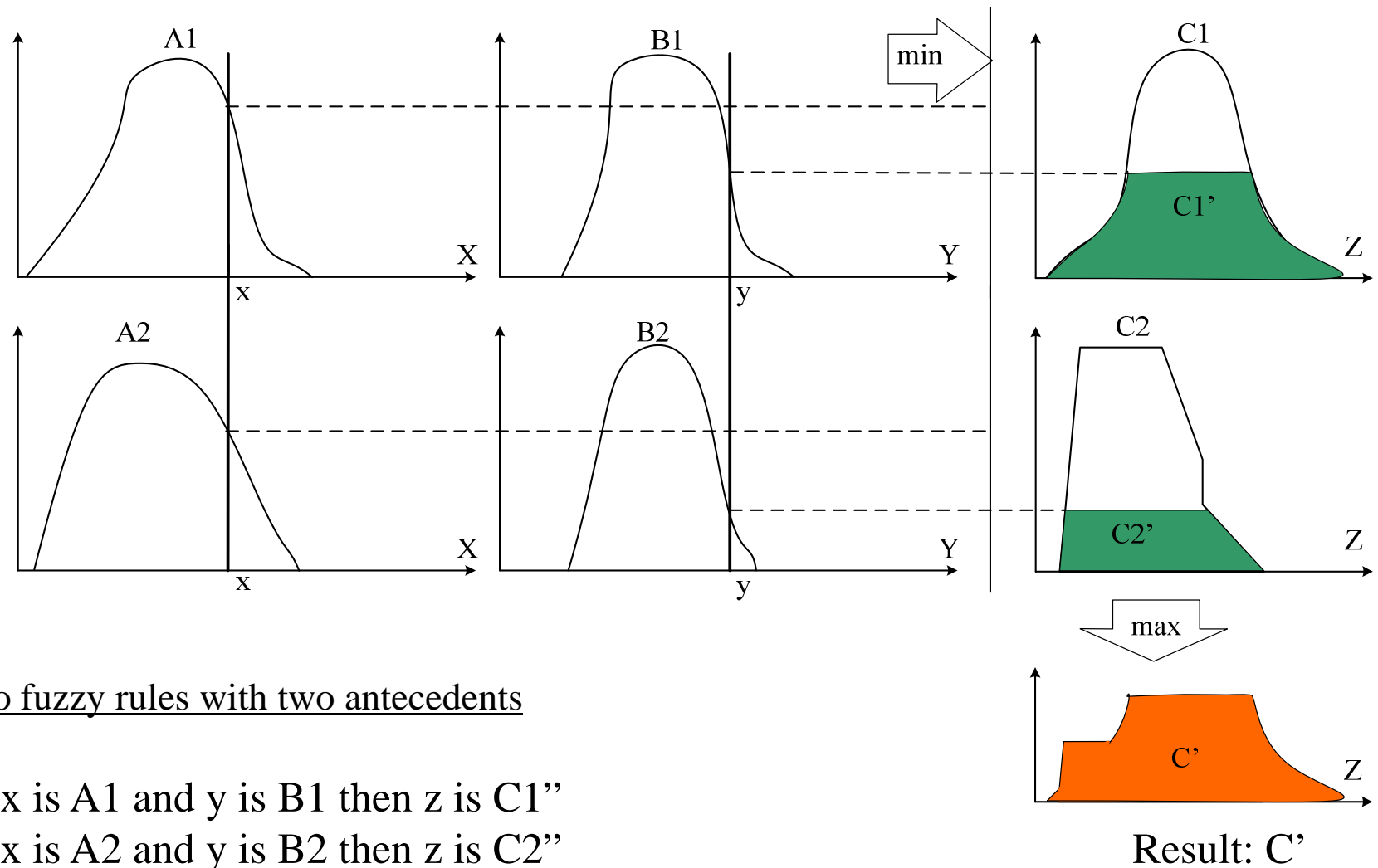
Defuzzyfication Schemes

Associate a numeric value
→ output of the fuzzy inference machine



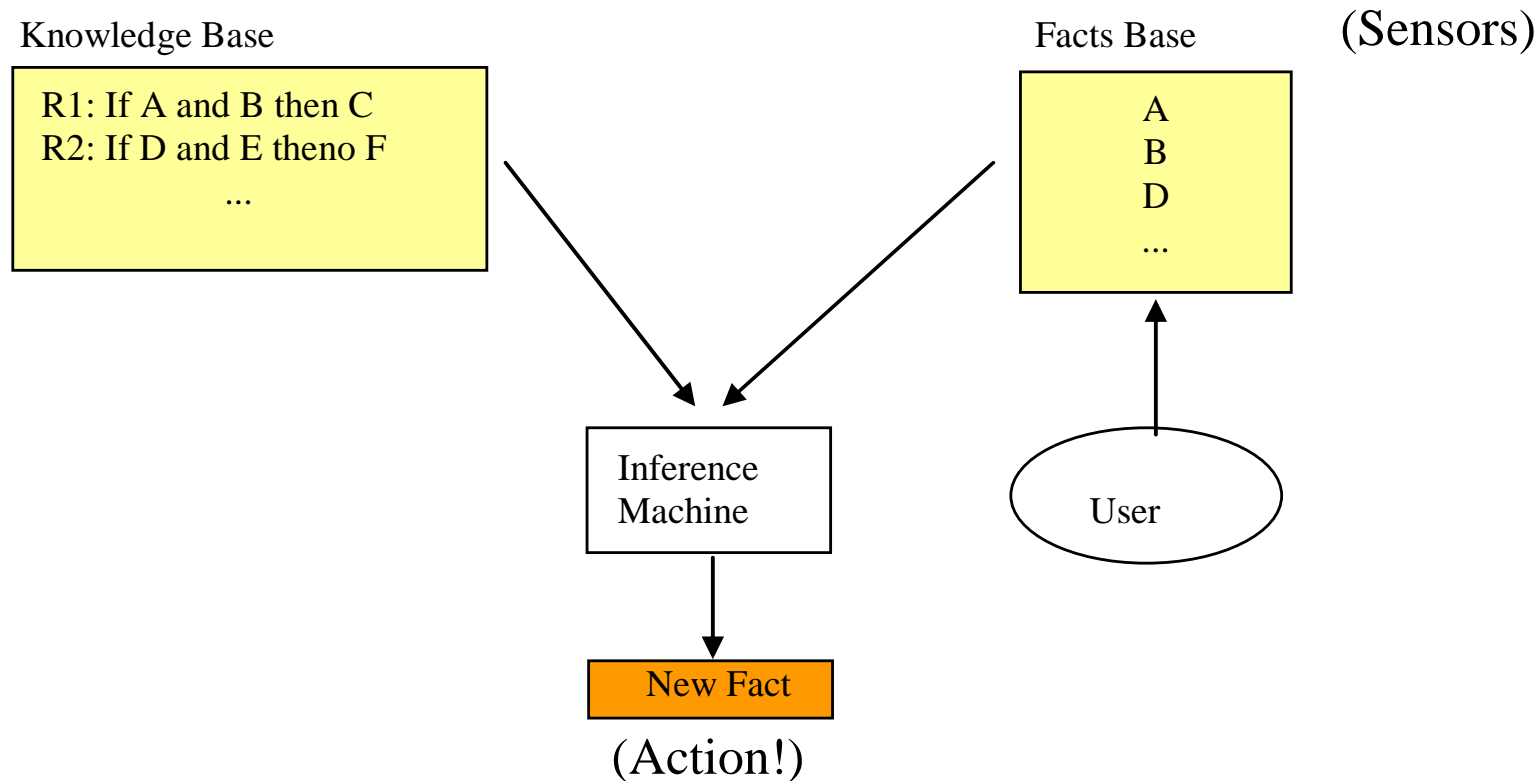
Fuzzy Inference with exact A' (“crisp”)

- Mamdani Model



Fuzzy Inference Systems

- Fuzzy systems are knowledge-based systems (like Expert Systems).



Fuzzy Inference Machine

The fuzzy inference machine follows these steps to obtain the Inference Result given a set of facts:

1. facts with premises (antecedents)
2. compatibility degree of each rule
3. belief in each rule
4. aggregation

For the aggregation four methods are popular:

- a) Mamdani (Max-Min)
- b) Larsen
- c) Tsukamoto
- d) Takagi-Sugeno

Exemple: Fuzzy Control (revisited)

– Air Conditioneer

Knowledge Base

R1: If T is *High* and U is *Low* then P is *average*
R2: If T is *Low* and U is *High* then P is *low*

Facts Base

$T = 30^{\circ}\text{C}$
 $U = 20\%$
 $R = 22^{\circ}\text{C}$

Inference
Machine

$P = 47\%$

User



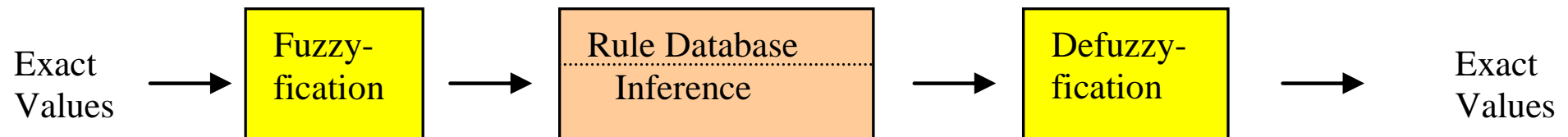
Window Air
Conditioneer

Supervisory
Control

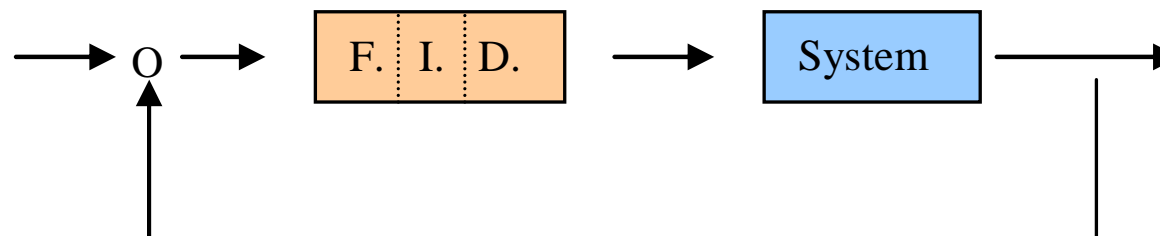


Interface with the Real World

•Fuzzyfication and Defuzzyfication



A feedback controller based on fuzzy logic (Intelligent Controller) would have the following structure, where F. I. D. means: Fuzzyfication, Inference and Defuzzyfication.



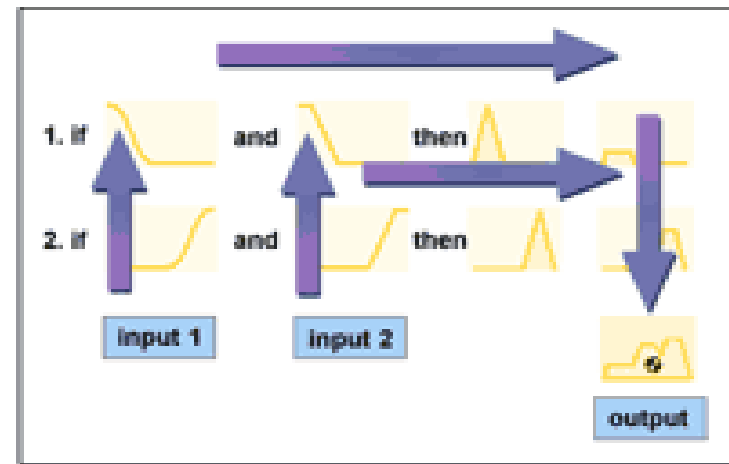
Xfuzzy

<http://www.imse.cnm.es/Xfuzzy/download.htm>

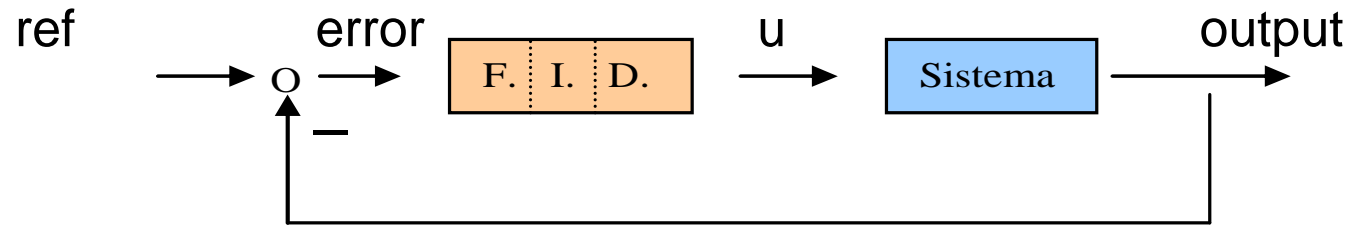
XFuzzy System for Unix developed by the
Instituto de Microelectrónica de Sevilla – Espanha

<http://www.mathworks.com> MatLab®

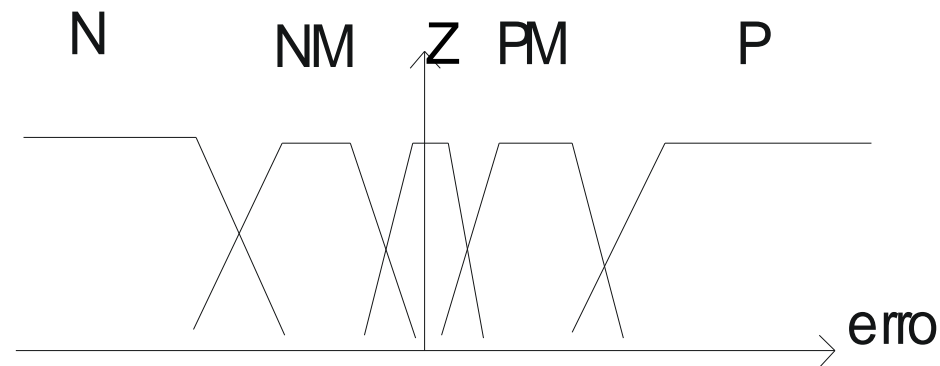
<http://www-rocq.inria.fr/scilab/> SciLab



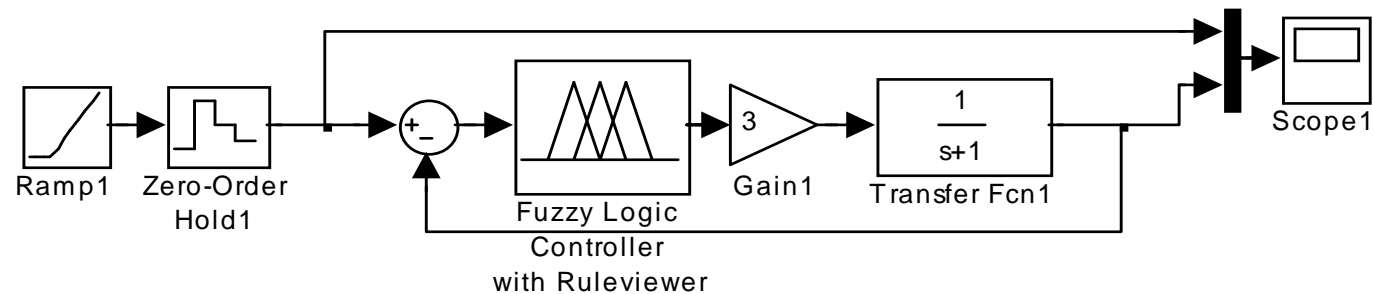
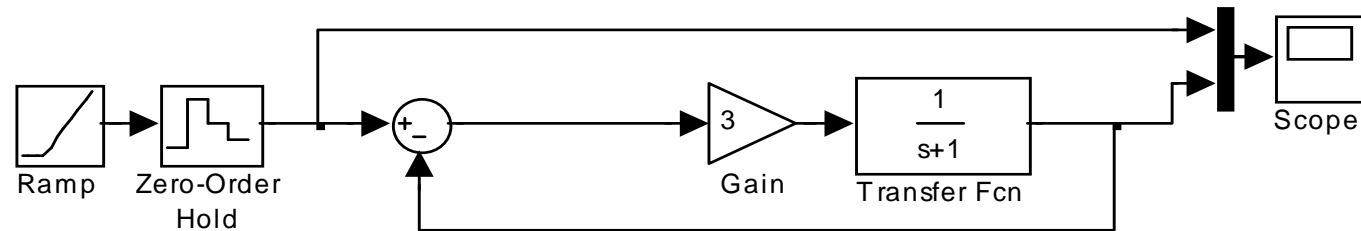
Ex: Fuzzy Proportional Control

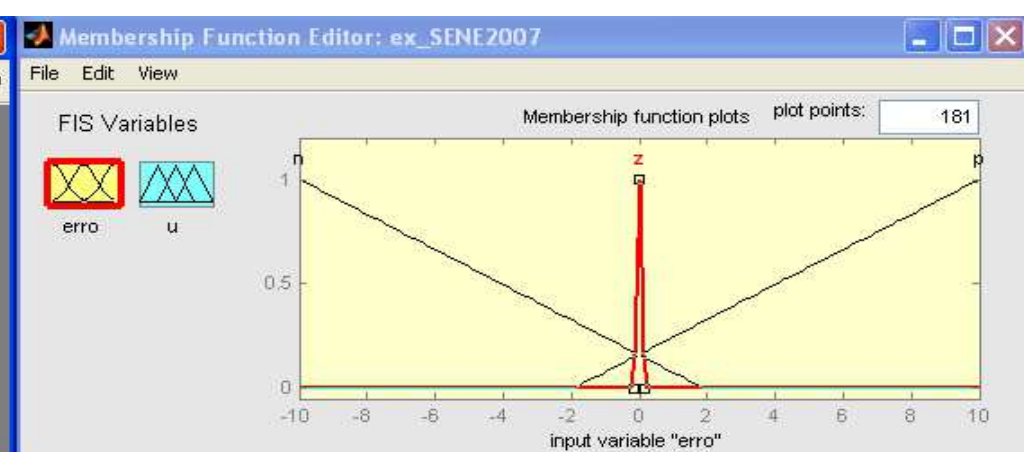
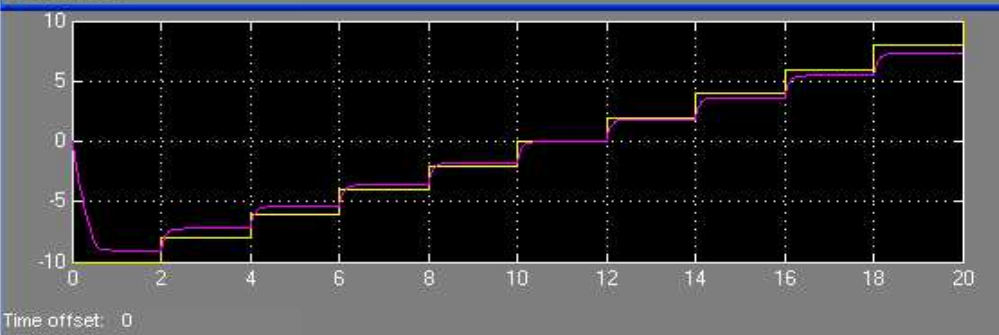
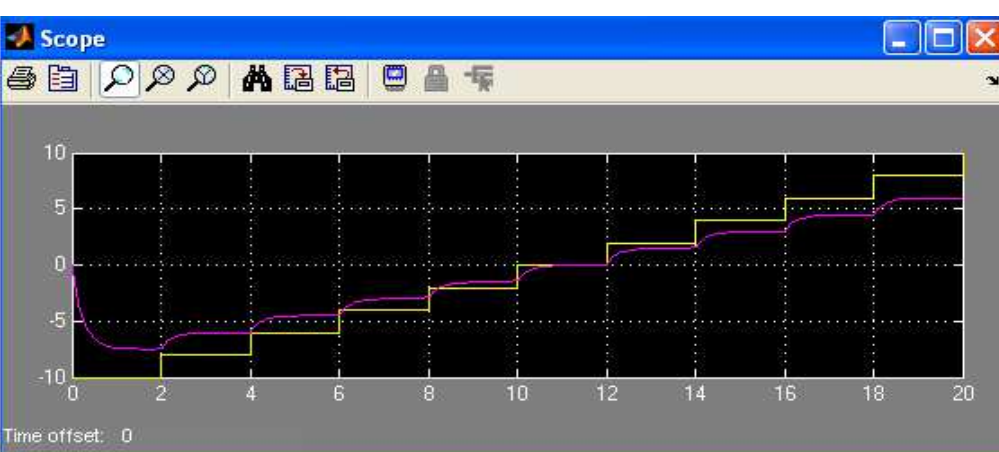


Closed Loop Control



Ex: Fuzzy Proportional Control





Current Variable

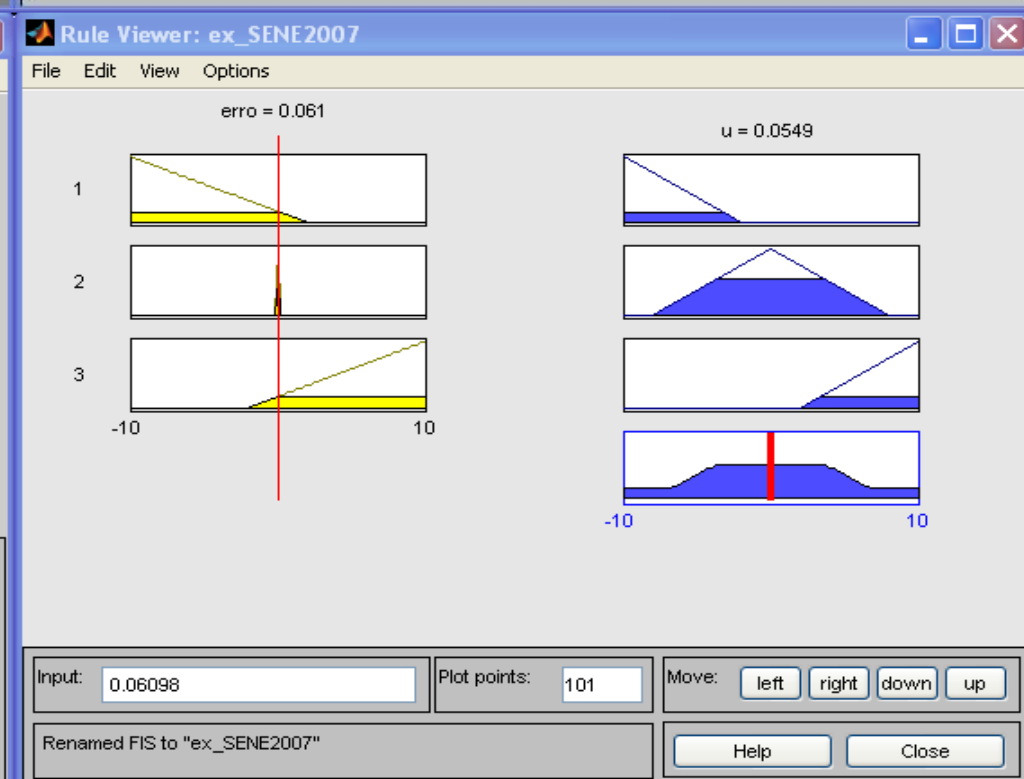
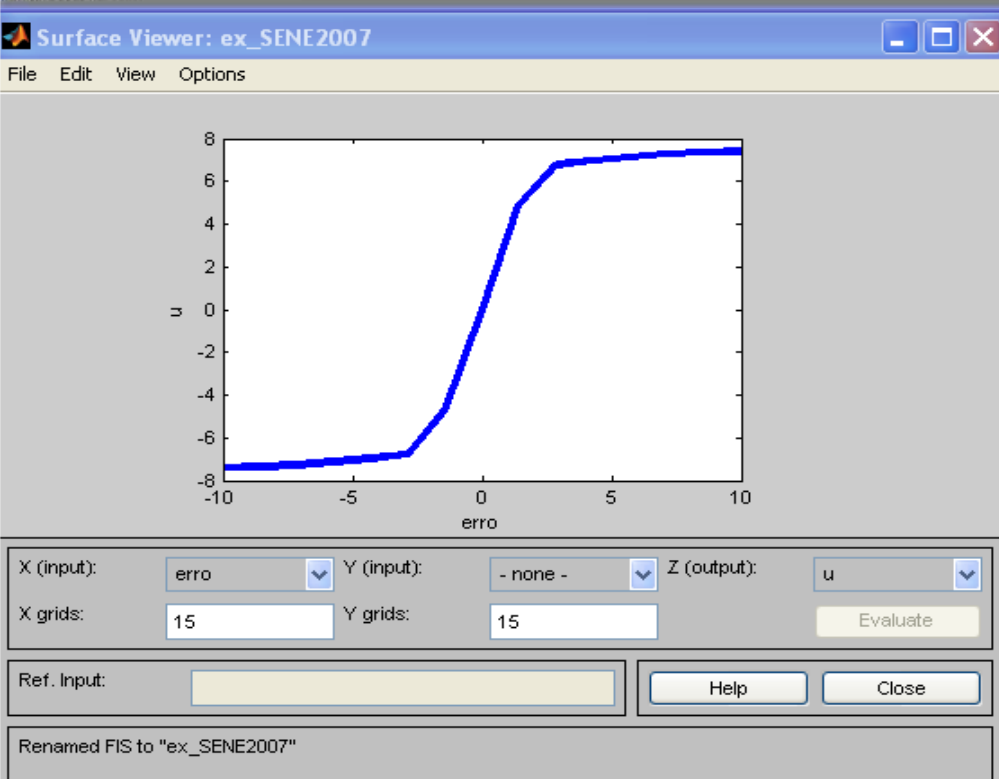
Name: erro
Type: input
Range: [-10 10]
Display Range: [-10 10]

Current Membership Function (click on MF to select)

Name: z
Type: trimf
Params: [-0.132 0 0.132]

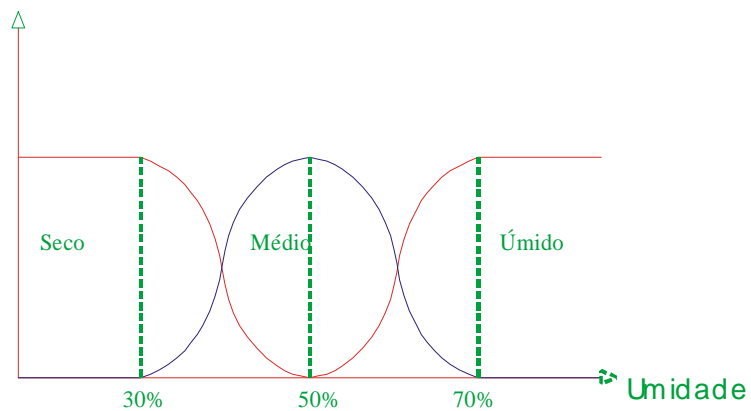
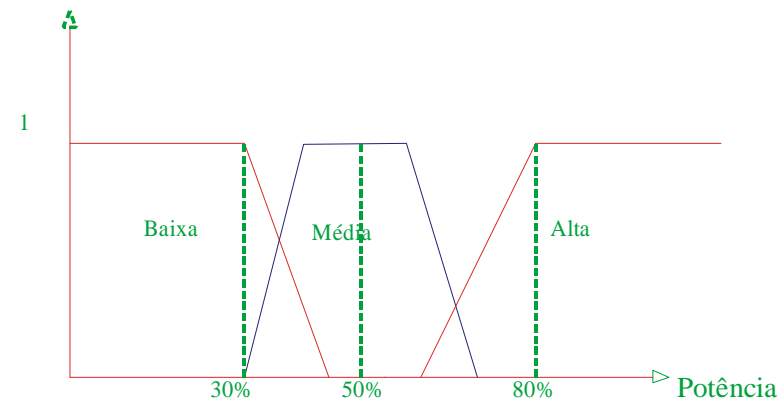
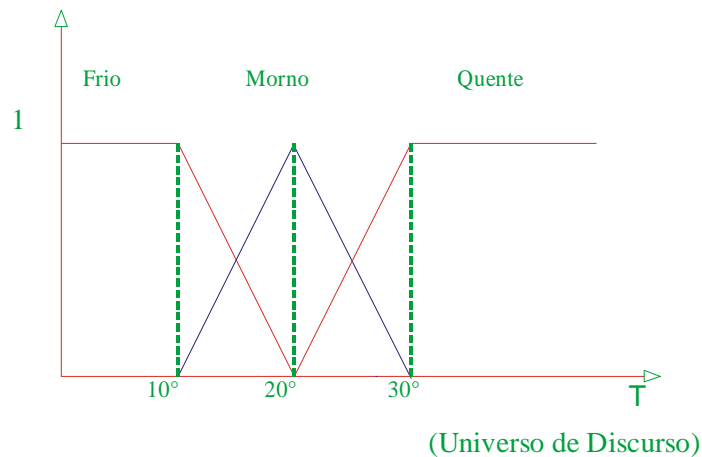
Help Close

Saved FIS "ex_SENE2007" to disk

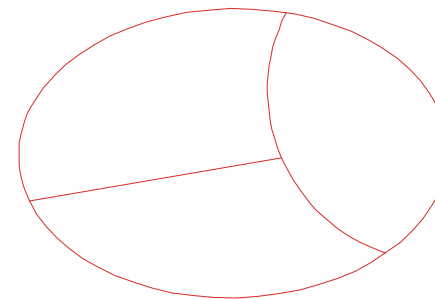


Ex: Multivariable *Fuzzy* Controller

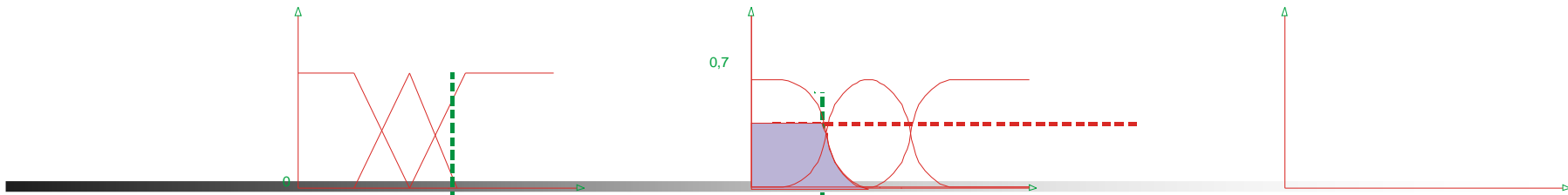
Air conditioning system for an office environment.



Conjunto Fuzzy
Partição do Universo de discurso



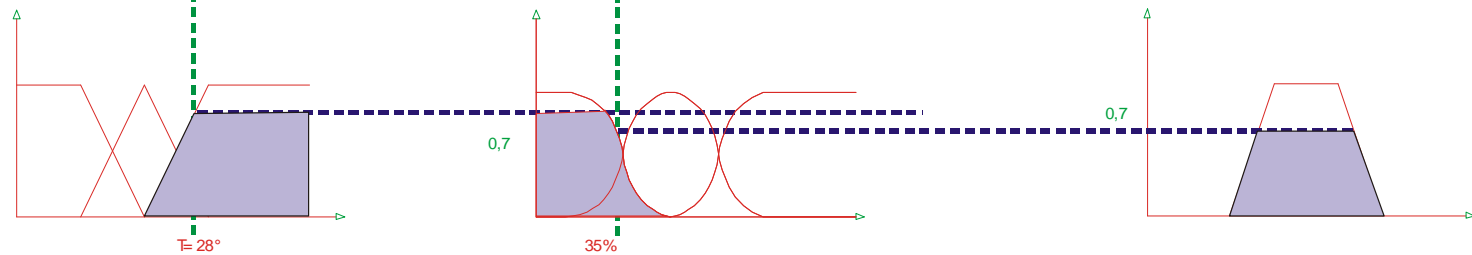
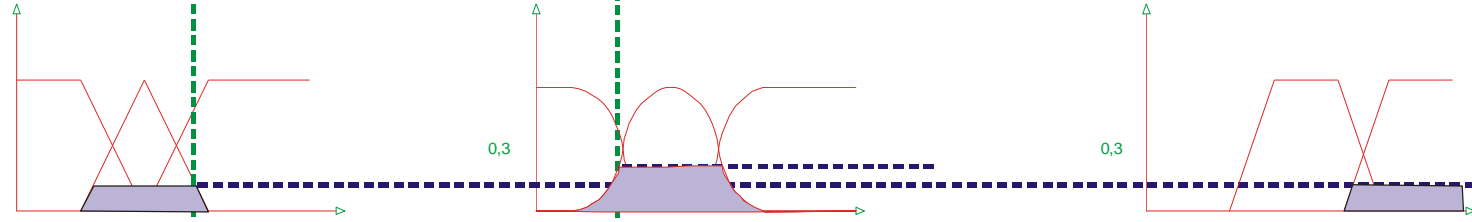
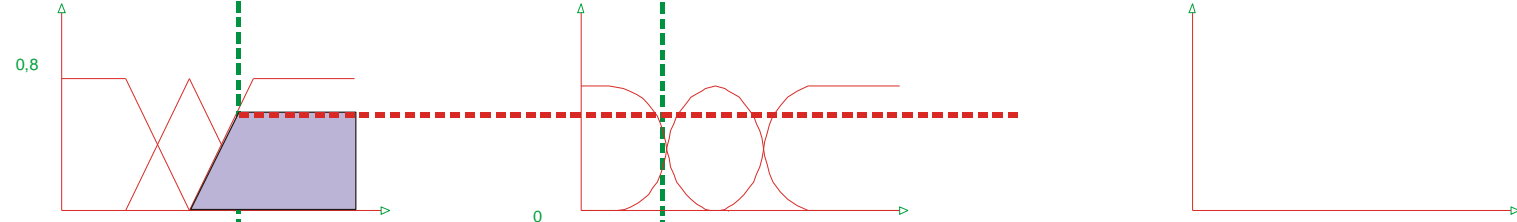
Membership functions used for the temperature control.



Fuzzy Inference:

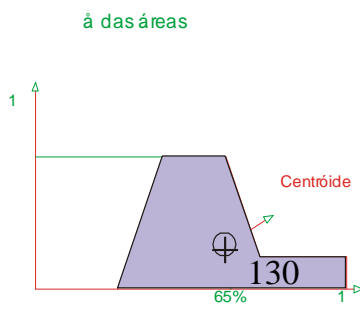
Temperature is 28° C
Relative humidity is 35%

The calculated Power is 65%.



Rule Basis

- If T is cold and U is dry then P is low
- If T is hot and U is humid then P is high
- If T is warm and U is average then P is high
- If T is hot and U is dry then P is average



Rule Editor: ex2_SENE2007

File Edit View Options

1. If (T is frio) and (U is seco) then (P is baixa) (1)
 2. If (T is quente) and (U is umido) then (P is alta) (1)
 3. If (T is conforto) and (U is medio) then (P is alta) (1)
 4. If (T is quente) and (U is seco) then (P is media) (1)

If T is and U is Then P is

frio seco baixa
 conforto medio media
 quente umido alta
 none none

not not not

Connection: or and

Weight: 1

Delete rule Add rule Change rule

The rule is added

Help Close

Membership Function Editor: ex2_SENE2007

File Edit View

FIS Variables

Membership function plots plot points: 181

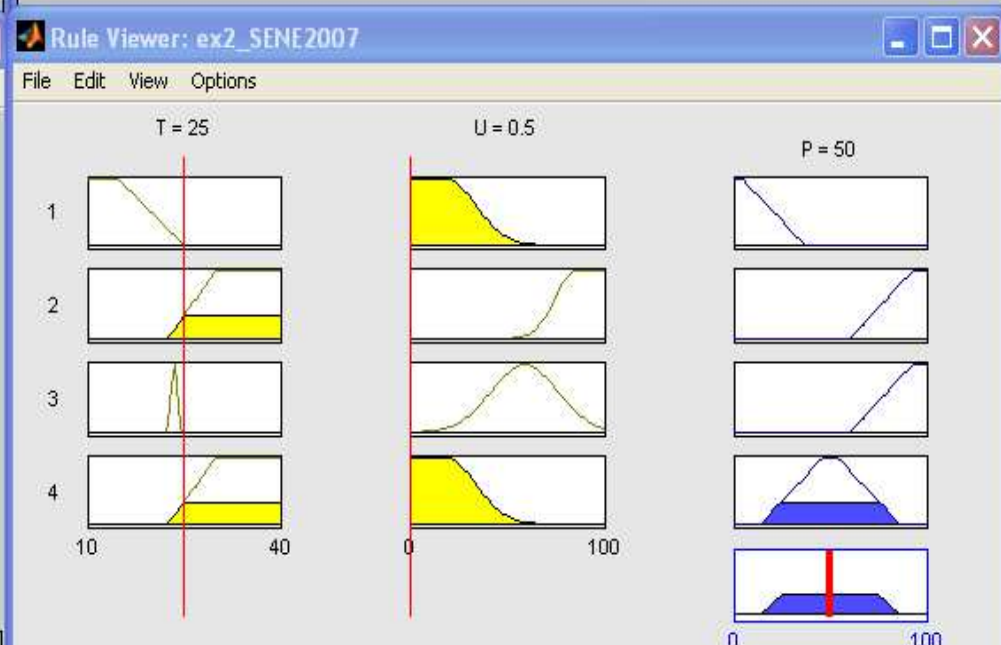
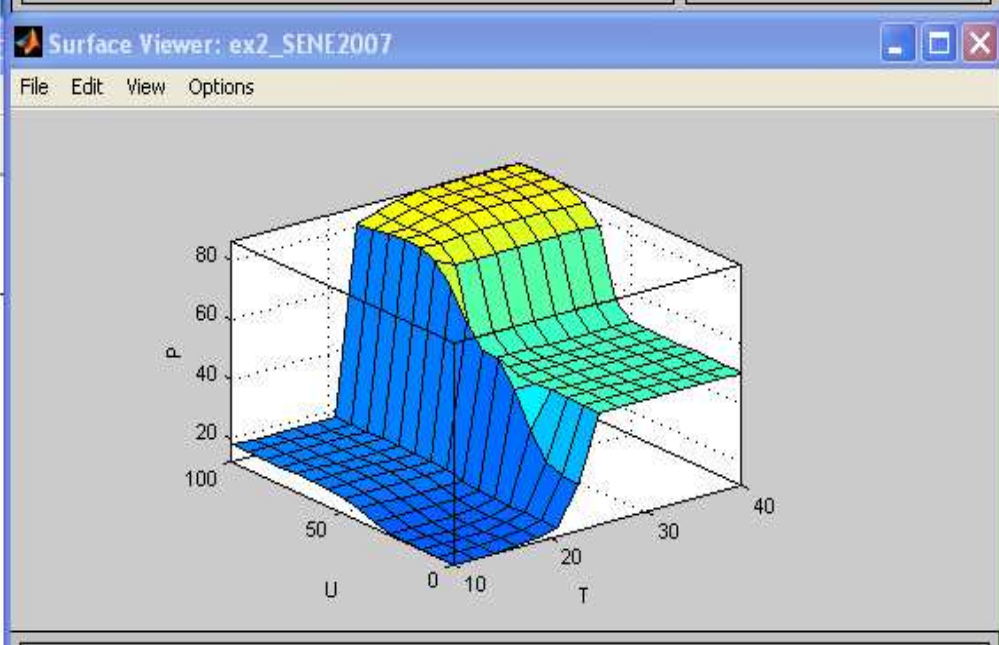
frio conforto quente

input variable "T"

Current Variable: Name T, Type input, Range [10 40], Display Range [10 40]

Current Membership Function (click on MF to select): Name conforto, Type trimf, Params [22.5 23.5 24.5]

Ready

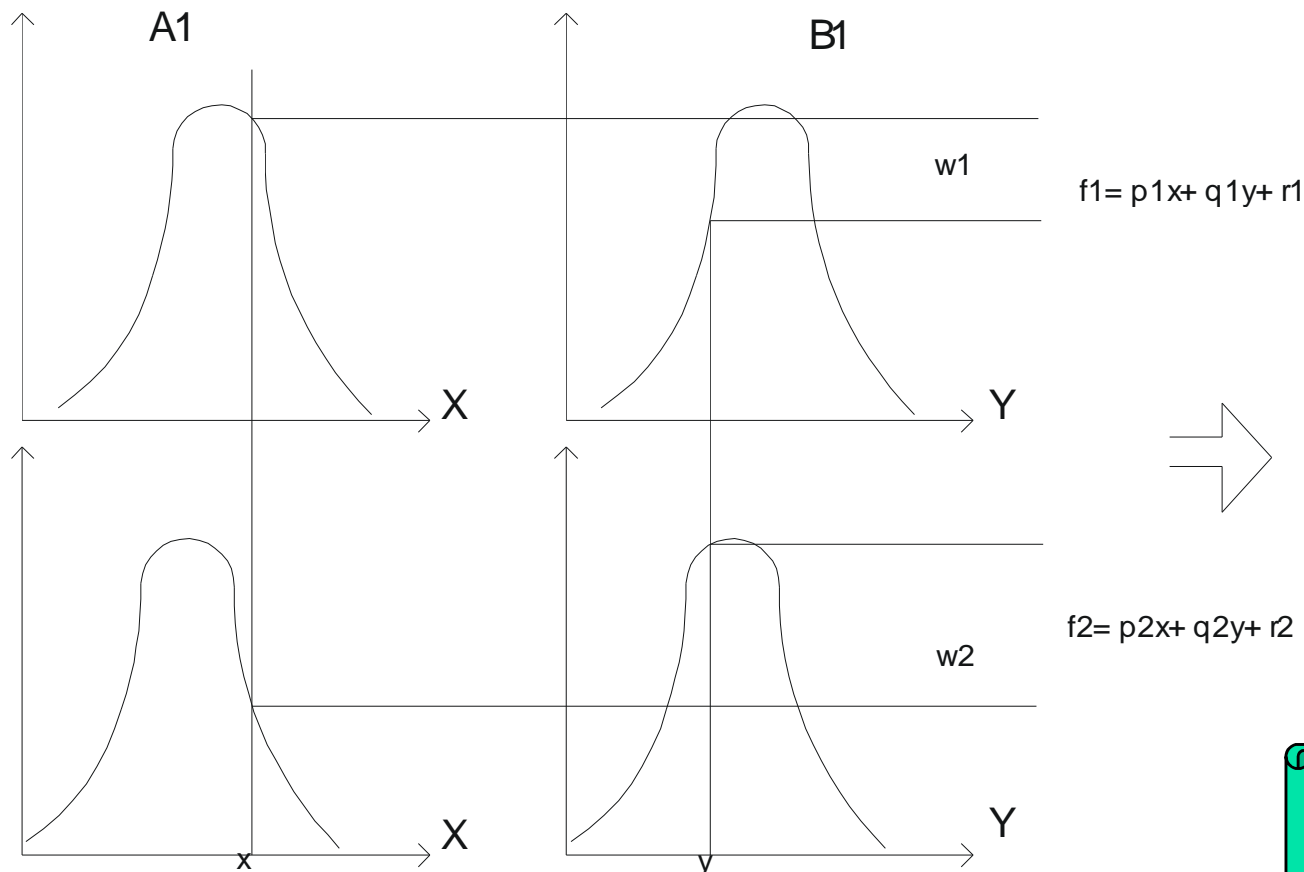


1st Order Sugeno Fuzzy Inference System

R_1 : IF x is A_1 AND y is B_1 THEN $f_1 = p_1x + q_1y + r_1$
 R_2 : IF x is A_2 AND y is B_2 THNE $f_2 = p_2x + q_2y + r_2$

Consequent:
linear combination
of the inputs

(p_i, q_i, r_i) instead of output M.F.

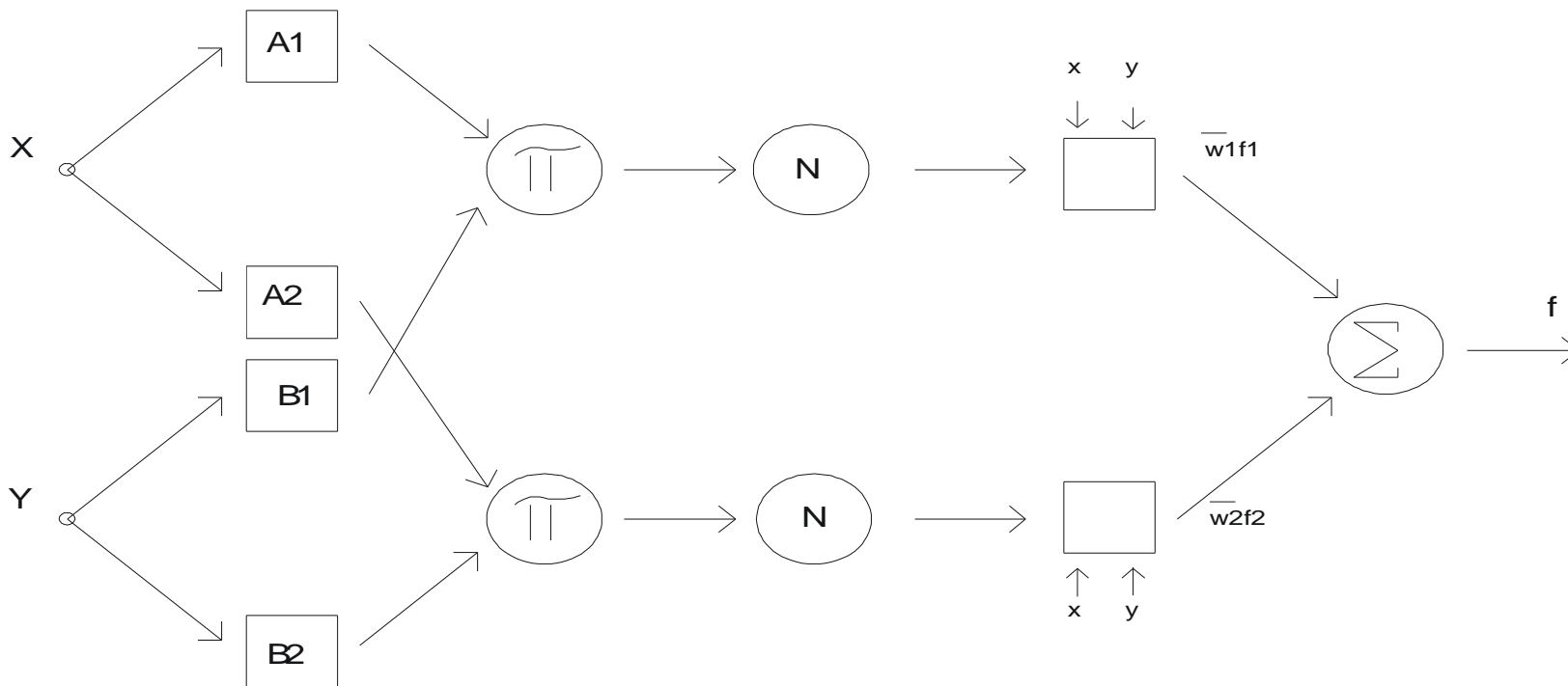


$$\begin{aligned}
 f &= \frac{w_1 f_1 + w_2 f_2}{w_1 + w_2} \\
 &= \bar{w}_1 f_1 + \bar{w}_2 f_2
 \end{aligned}$$

Degrees of compatibility (w_1, w_2)
weigh the rule interpolation

Adaptive Neuro Fuzzy Inference System (ANFIS)

R_1 : IF x is A_1 AND y is B_1 THEN $f_1 = p_1x + q_1y + r_1$
 R_2 : IF x is A_2 AND y is B_2 THNE $f_2 = p_2x + q_2y + r_2$



Adaptive feed-forward network for the Sugeno fuzzy model

→ adaptive block
 → fixed block

ANFIS Layers

Layer 1: Adaptive Nodes

$$o_{1,i} = \mu_{A_i}(x), \text{ for } i = 1, 2.$$

$$o_{1,i} = \mu_{B_{i-2}}(y), \text{ for } i = 3, 4.$$

A_i – generalized bell function

$$\mu_{A_i}(x) = \frac{1}{1 + [(x-c_i)^2/(a_i)^2]^{b_i}}$$

$\{a_i, b_i, c_i\}$ set of premise parameters

Layer 2: Fixed Nodes

$$o_{2,i} = \omega_i = \mu_{A_i}(x) \times \mu_{B_i}(y), \quad i = 1, 2$$

↓
T-Norm

Layer 3: Fixed Nodes

$$o_{3,i} = \omega_i \text{ (m\u00e9dio)} = \frac{\omega_i}{\omega_1 + \omega_2}, \quad i = 1, 2$$

Layer 4: Adaptive Nodes

$$o_{4,i} = \frac{\omega_i}{\omega_1 + \omega_2} \cdot f_i = \frac{\omega_i}{\omega_1 + \omega_2} \cdot (p_i x + q_i y + r_i), \quad i = 1, 2$$

$\{p_i, q_i, r_i\}$ set of the consequent parameters

Layer 5: Fixed Node

$$o_{5,i} = \sum_i \frac{\omega_i}{\omega_1 + \omega_2} \cdot f_i = \frac{\sum_i \omega_i f_i}{\sum_i \omega_i}$$

ANFIS Hybrid Learning

1 – Fix the premiss parameters

⇒ Output is a linear combination of the consequence parameters

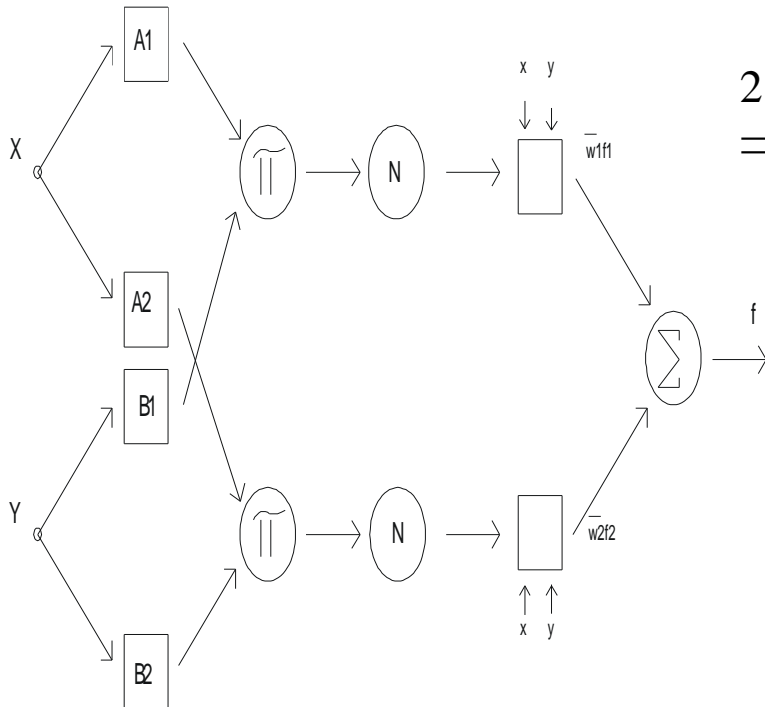
$$f = \frac{\omega_1}{\omega_1 + \omega_2} \cdot f_1 + \frac{\omega_2}{\omega_1 + \omega_2} \cdot f_2$$

$$= \frac{1}{\omega_1 + \omega_2} \cdot (\omega_1 x p_1 + \omega_1 y q_1 + \omega_1 r_1 + \omega_2 x p_2 + \omega_2 y q_2 + \omega_2 r_2)$$

⇒ Identify consequence parameters using Least Mean Squares method.

2 – Backpropagation of the error signals to adapt the premiss parameters

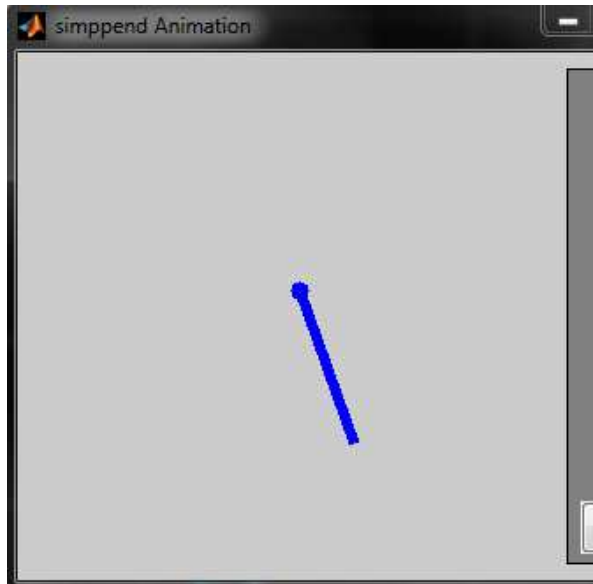
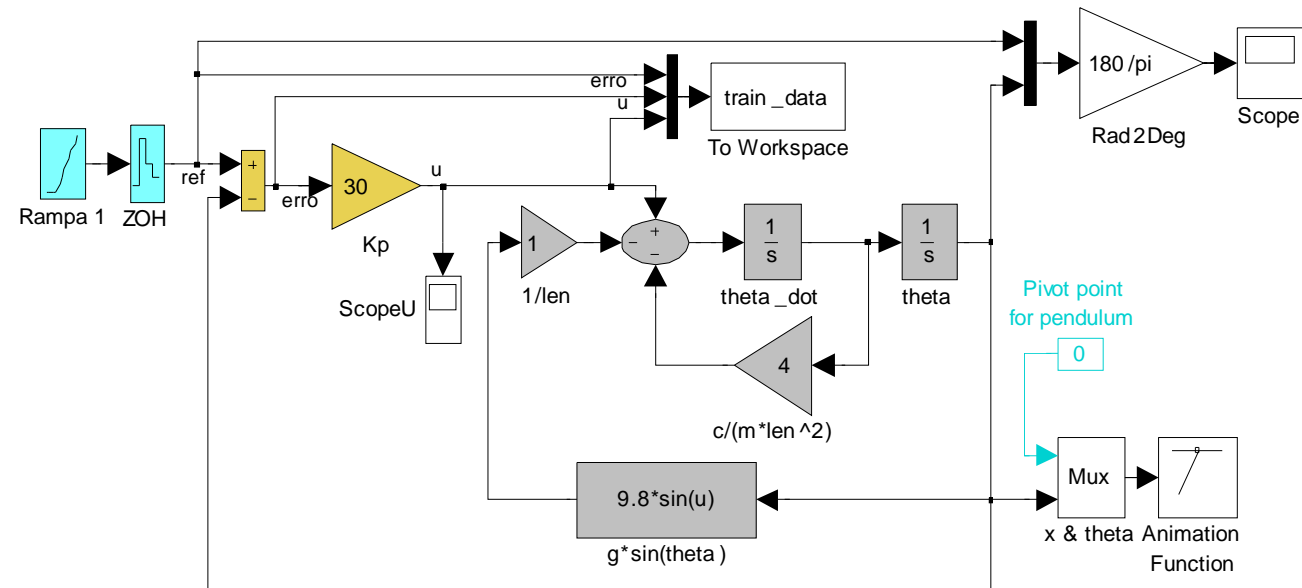
⇒ Gradient descent method.



	Forward Step	Backward Step
Premisse Parameters	Fixed	Gradiente descent
Consequence Parameters	LMS Estimate	Fixed
Signals	Output nodes	Error Signals

ANFIS

Adaptive Neuro Fuzzy Inference System

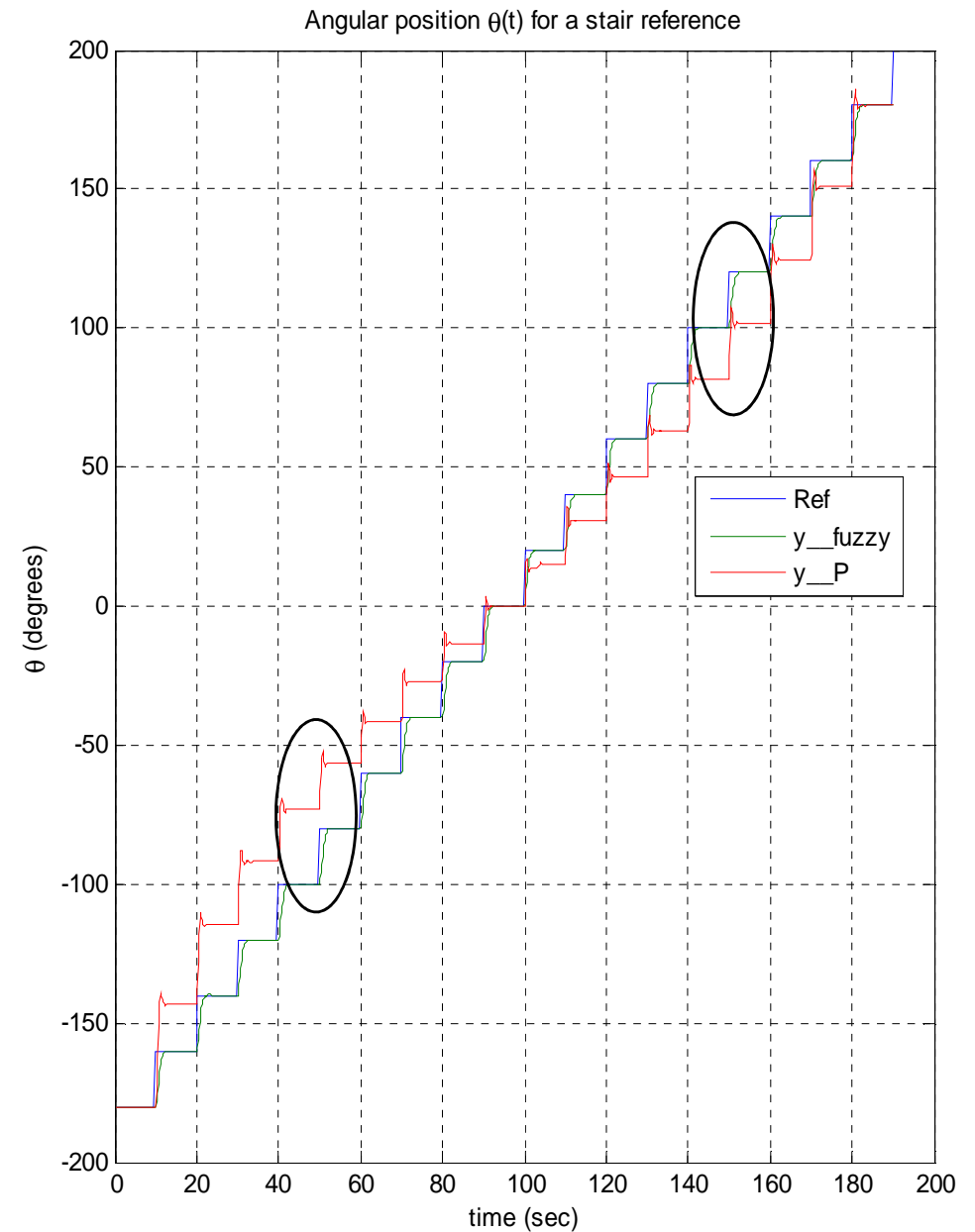
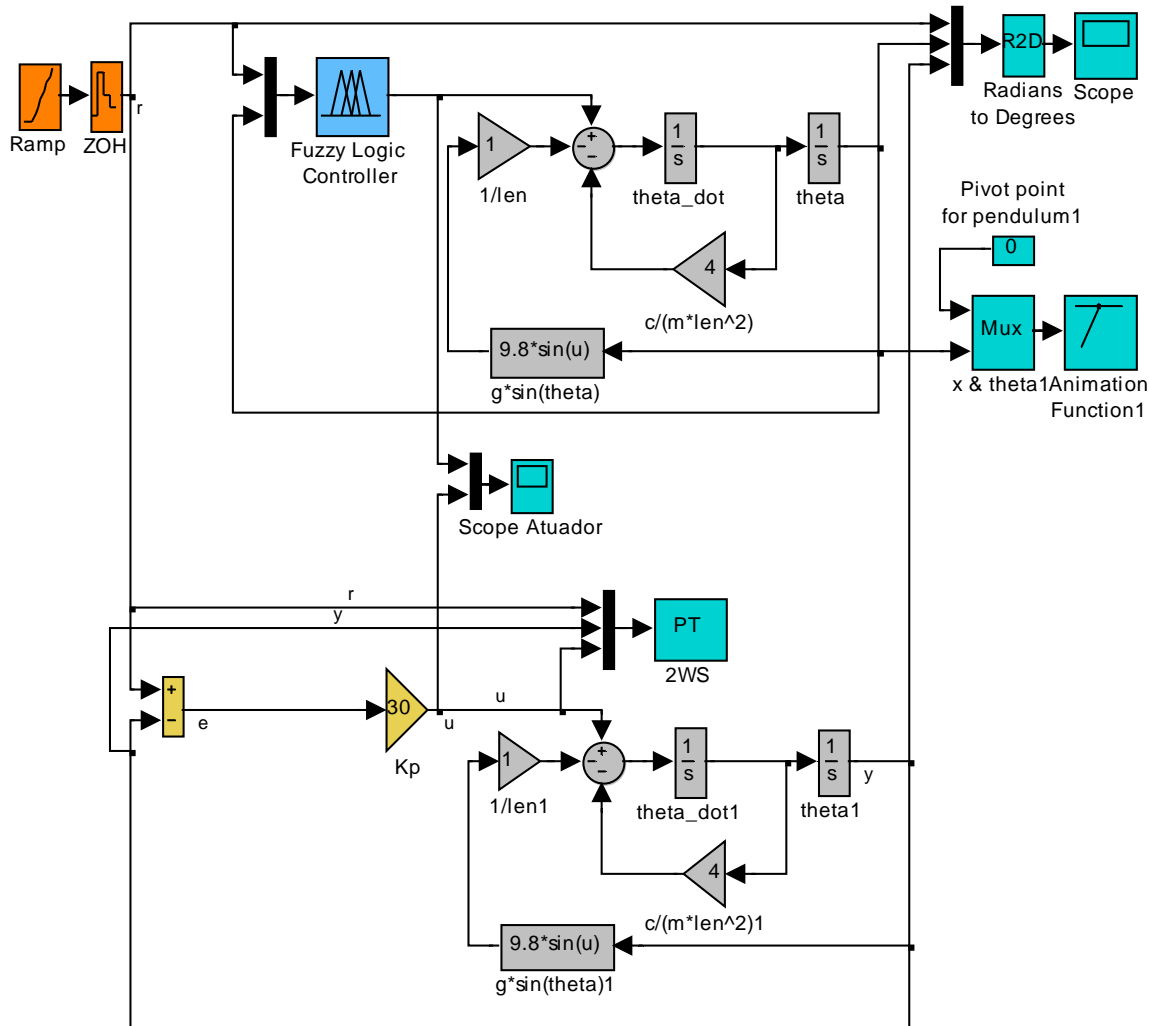


$$\ddot{\theta} = -\frac{c}{ml^2} \dot{\theta} - \frac{g \sin(\theta)}{l} + u$$

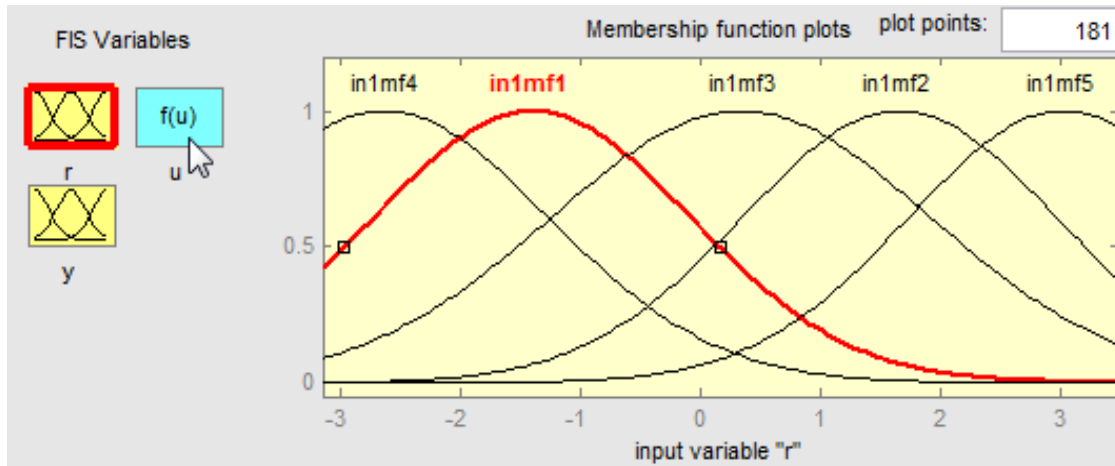
$$\ddot{\theta} + 4\dot{\theta} + 9.8 \sin(\theta) = u$$

$$\text{Control law } u = 30(r - \theta)$$

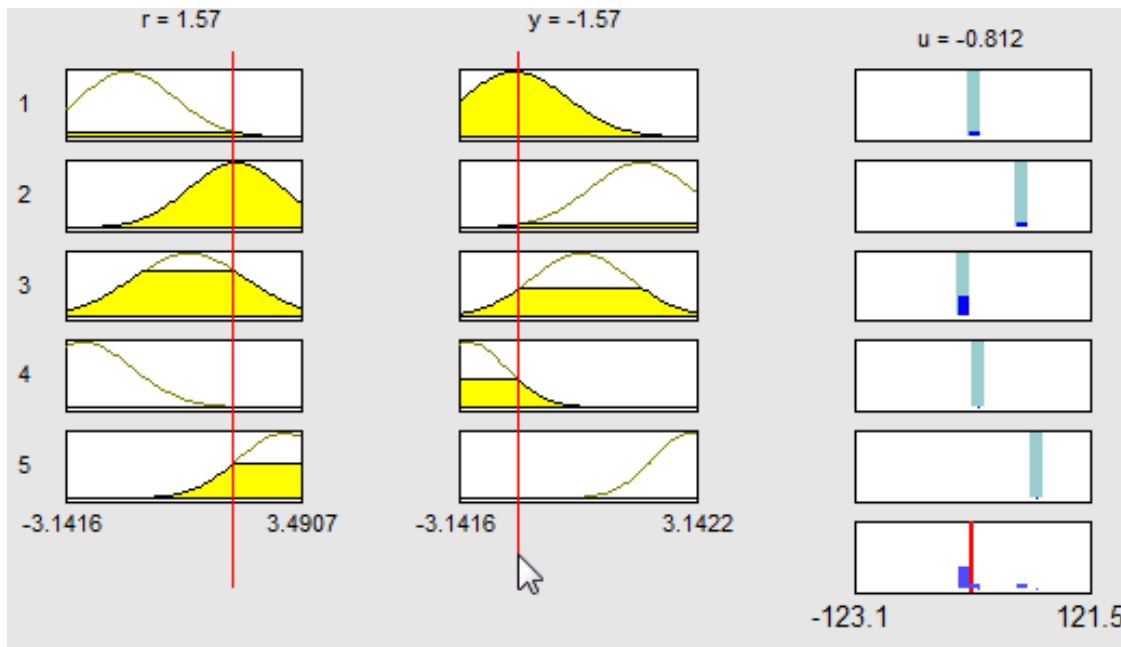
ANFIS x P-Controller



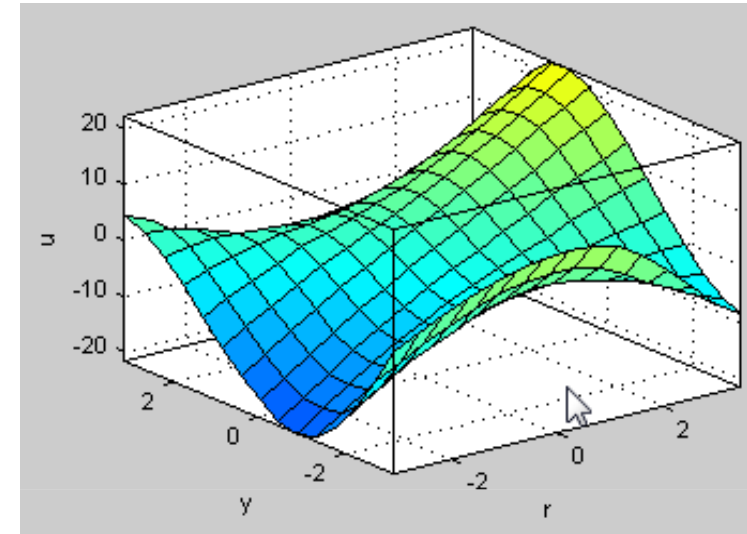
ANFIS Controller



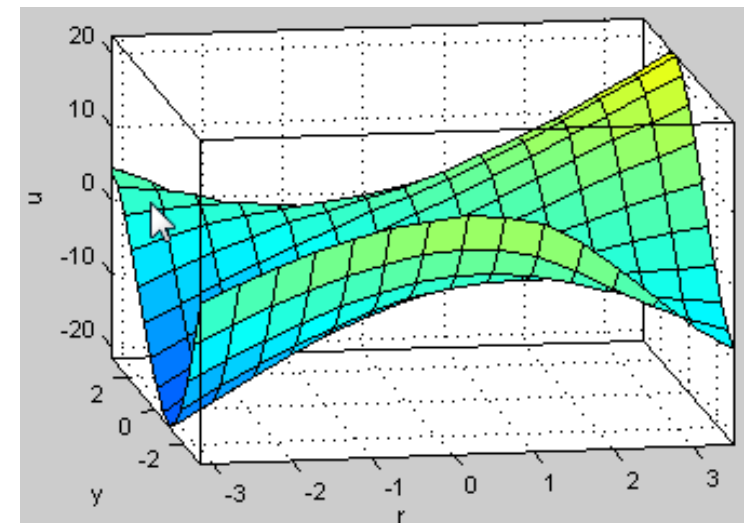
Membership functions of r (reference)



Rule Viewer

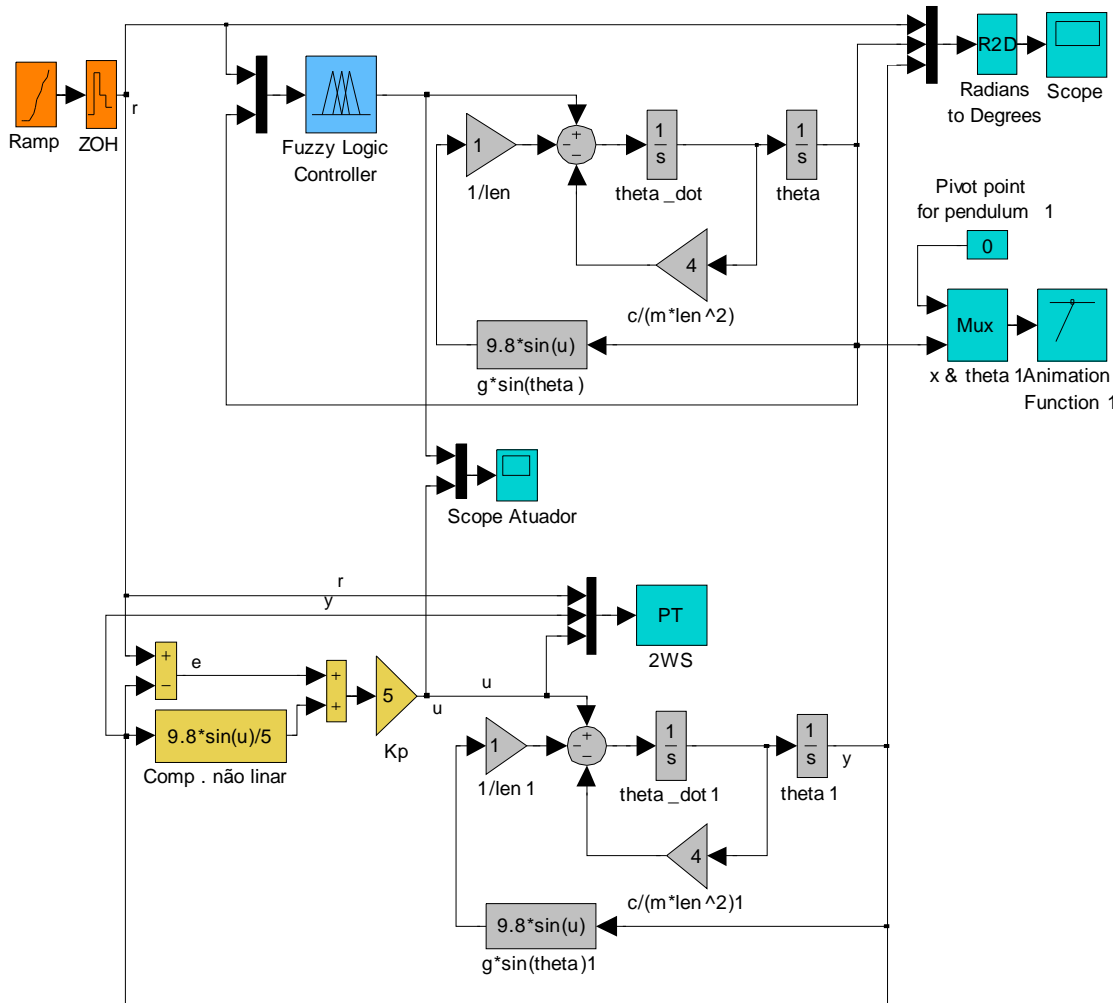


View of the *non-linear* control surface



Another view of the *non-linear* control surface

ANFIS – Who is the Expert?



Expert Knowledge

Linear behavior if the control signal, u , can cancel the non-linear dynamics of the process.

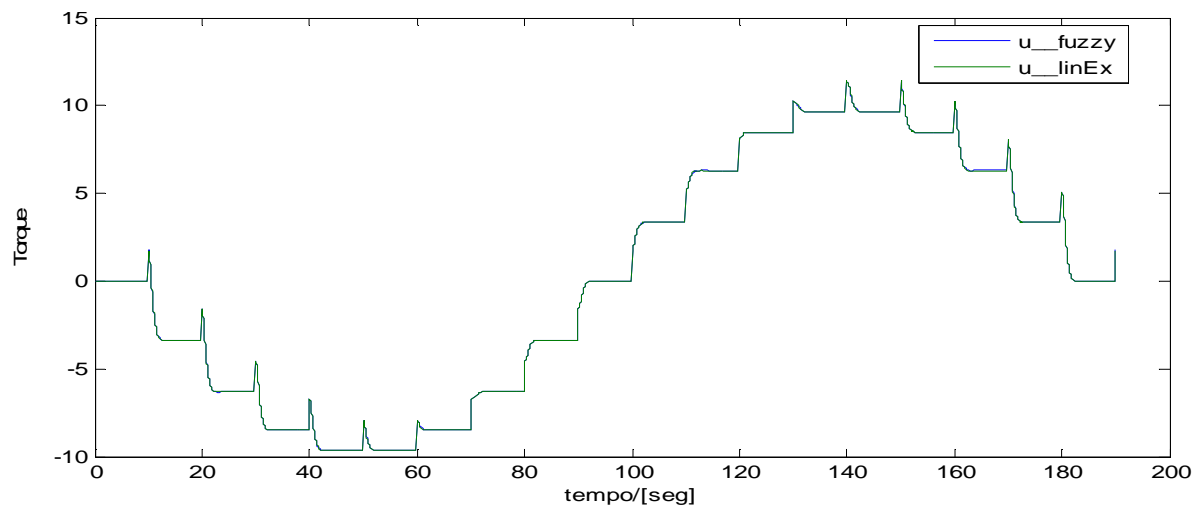
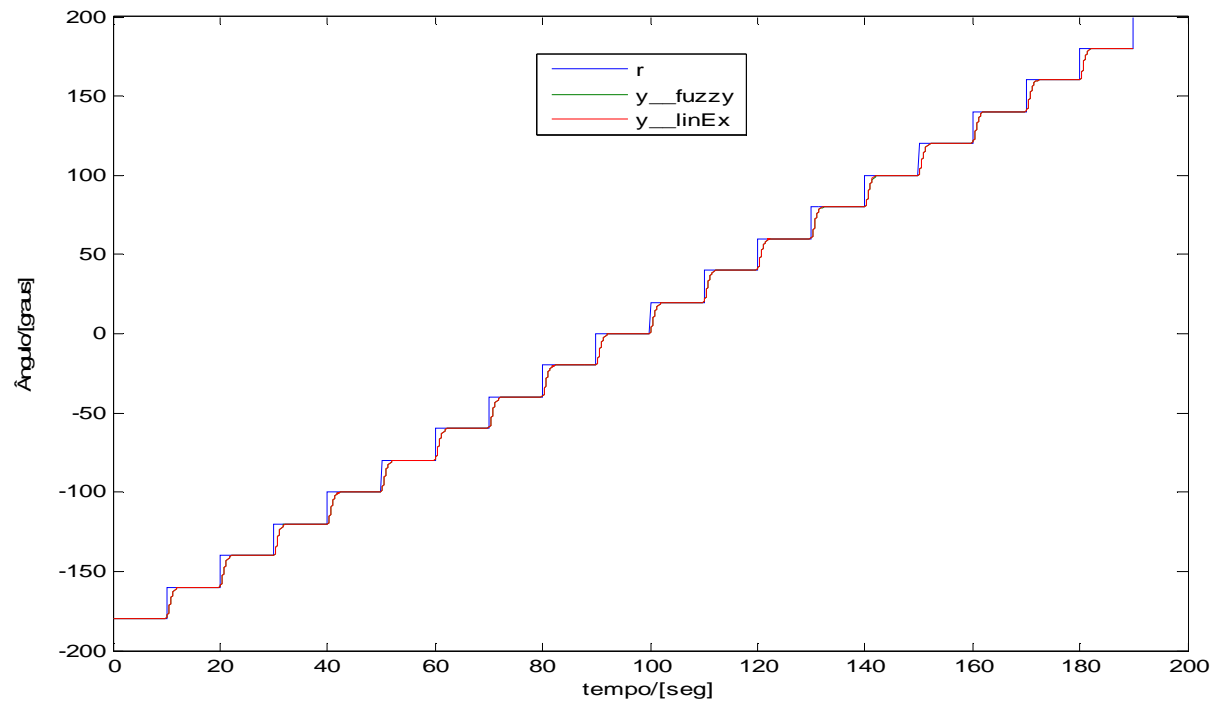
→ “Exact Linearization” (E.L.)
(not only operating point)

$$\ddot{\theta} + 4\dot{\theta} + 9.8 \sin(\theta) = u$$

$$P - \text{Control law} \quad u = K_p (r - \theta)$$

$$E.L. - \text{Control law} \quad u = K_p (r - \theta - 9.8 \sin(\theta))$$

ANFIS – Controller



Rules Trained by an ANN!

- You can explain and add new rules (*fuzzy*)
- You can train with real data (ANN)
- Drawback
more parameters...