

# Control Of An Inverted Pendulum With the Help Of A Piece-wise Affine Fuzzy Controller

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**Abstract :** A fuzzy controller, a conventional variable structure controller, and a region-wise affine fuzzy controller are compared for the control of an inverted pendulum. The region-wise affine controller is designed via a conventional fuzzy controller with a linear defuzzification algorithm; the input and output membership functions parameters are chosen in order to ensure the stability of the linearised system.

**Keywords :** fuzzy control, stability, fuzzy controller design

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## 1. Introduction

Experimental robustness properties of fuzzy controllers remain theoretically difficult to prove and their synthesis is still an open problem. The non-linear structure of the final

controller is derived from all controllers at the different stages of fuzzy control [1], particularly from common defuzzification methods (such as Centre of Area). In general, fuzzy controllers have a region-wise structure given the partition of its input space by the fuzzification stage. Local controls designed in these regions are then combined into sets to make up the final global control. A partition of the state space can be found for which the controller has region-wise constant parameters [3], [4]. Moreover, each fuzzy controller tuning parameter (i.e. the shapes and the values of input or output variables membership functions) influences the values of parameters in several regions at the same time. In the particular case of a switching line separating the phase plane into one region where the control is positive whereas in the other it is negative, the fuzzy controller may be seen as a variable structure controller [5] [6] [7] [8]. This kind of a fuzzy controller can be assimilated to a variable structure controller with boundary layer such as in [9], for which stability theorems exist, but with a non-linear switching surface [7] [8].

With the use of trapezoidal input membership functions and appropriate composition and inference methods, it will be shown that it is possible to obtain rule membership functions which are region-wise affine functions of the controller input variable. We propose a linear defuzzification algorithm that keeps this region-wise affine structure and yields a piece-wise affine controller. A particular and systematic parameter tuning method will be given which allows to turn this controller into a variable structure-like controller. We will compare this region-wise affine controller with a Fuzzy and Variable Structure Controller

through the application to an inverted pendulum control [10].

In the first Section, the equations of the inverted pendulum system will be recalled. In the second Section, the classical fuzzy and variable structure controllers, to be used in the system control, will be presented. A fuzzy controller of which output is an affine function of its input variables will then be designed, and a condition on the fuzzy controller parameters values will be set so that the resulting control should be piece-wise linear and stable when used to control linear systems. In the third Section, simulations will be referred for the three controllers and a short discussion on the piece-wise affine controller behaviour will be undertaken.

## 2. Inverted Pendulum System

The system to be controlled is the classical inverted pendulum system described in Figure 1 [10].

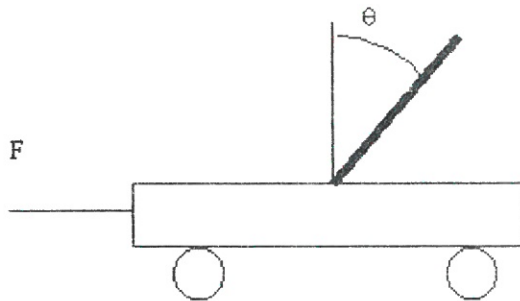


Figure 1. Inverted Pendulum

A force  $F$  is applied to the cart. The goal is to stabilise the rod to a position  $\theta = 0$  from an initial position  $\theta_0$ , following a reference trajectory

$$\theta_d = 0.5\theta_0(1 + \cos(\pi t)). \quad (1)$$

The length of the rod is  $2l$ , with  $l = 1$  m, the mass of the rod is  $m = 0.1$  kg, the mass of the cart is  $m_c = 1$  kg.

The angle follows the equation:

$$\ddot{\theta} = f(\vartheta) + g(\vartheta)F, \quad (2)$$

where  $F$  is the control, and

$$f(\theta) = \left( g \sin(\theta) - \frac{ml}{m+m_c} \cos(\theta) \sin(\theta) \dot{\theta}^2 \right) / \left( l \left( \frac{4}{3} - \frac{m}{m+m_c} \cos^2(\theta) \right) \right), \quad (3)$$

$$g(\theta) = -\cos(\theta) / \left( l \left( \frac{4}{3} - \frac{m}{m+m_c} \cos^2(\theta) \right) \right). \quad (4)$$

This model can be linearised around the equilibrium point using the model

$$\ddot{\theta} = \tilde{f}(\vartheta) + \tilde{g}(\vartheta)F = \left( g\theta - \frac{m}{m+m_c} F \right) / \left( l \left( \frac{4}{3} - \frac{m}{m+m_c} \right) \right) \quad (5)$$

which can take the form :

$$\dot{x} = Ax + BF, \quad (6)$$

We will define the control error as

$$\varepsilon = \vartheta - \theta_d \text{ and the vector } x \text{ as } x = \begin{pmatrix} \dot{\varepsilon} \\ \varepsilon \end{pmatrix}.$$

## 3. Fuzzy and Variable Structure Controllers

### A. Fuzzy Control of the Inverted Pendulum

The control error  $\varepsilon$  and its derivative  $\dot{\varepsilon}$  are the controller inputs.

The proposed fuzzy controller [1] follows the linguistic rules  $R_{j,k}$  such as:

$$\text{if } \varepsilon \text{ is } A_j \text{ AND } \dot{\varepsilon} \text{ is } B_k \text{ THEN } u \text{ is } U_{j,k},$$

where  $A_j$  and  $B_k$  are linguistic predicates. For the sake of simplicity,  $U_{j,k}$  will be a scalar number.

A membership function  $\mu_j(\varepsilon)$  to the predicate " $\varepsilon$  is  $A_j$ " and  $\mu_k(\dot{\varepsilon})$  of the predicate " $\dot{\varepsilon}$  is  $B_k$ " can be defined. The value of the membership functions to the predicate " $\varepsilon$  is  $A_j$  AND  $\dot{\varepsilon}$  is  $B_k$ " is defined as

$$\mu_{j,k}(\varepsilon, \dot{\varepsilon}) = \min(\mu_j(\varepsilon), \mu_k(\dot{\varepsilon})). \quad (7)$$

The output  $F$  is computed with the centroid method:

$$F(\varepsilon, \dot{\varepsilon}) = \frac{\sum \mu_{j,k} U_{j,k}}{\sum \mu_{j,k}} \quad (8)$$

Practically, 3 predicates (Negative (N), Zero (Z), Positive (P)) will be considered; the composition function AND will be the function  $\min(\cdot)$  as stated in formula (7). The trapezoidal membership functions are defined in Figure 2.

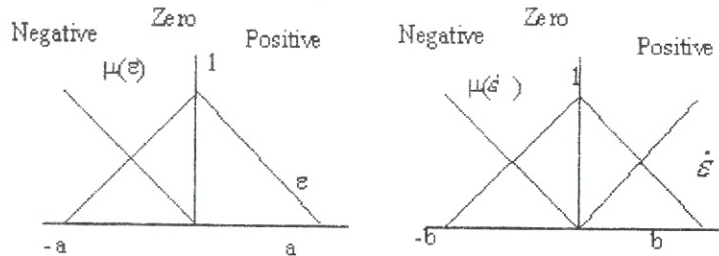


Figure 2. Input Membership Functions

The look-up table is defined in Table 1.

Table 1. Lookup Table and Rule Number (in parentheses) for the Fuzzy Controller Output

$\varepsilon \backslash \dot{\varepsilon}$	Negative	Zero	Positive
Positive	$F_{1,1} (R_{1,1})$	$F_{1,2} (R_{1,2})$	$F_{1,3} (R_{1,3})$
Zero	$F_{2,1} (R_{2,1})$	$F_{2,2} (R_{2,2})$	$F_{2,3} (R_{2,3})$
Negative	$F_{3,1} (R_{3,1})$	$F_{3,2} (R_{3,2})$	$F_{3,3} (R_{3,3})$

After some tuning trials, the following parameters are obtained:

$$a = 0.005, b = 0.02, F_{1,1} = -90, F_{1,2} = 6, F_{1,3} = 90, F_{2,1} = -30, F_{2,3} = 30, F_{3,1} = -90, F_{3,2} = -6, F_{3,3} = 90, F_{2,2} = 0.$$

### B. Sliding Mode Control of the Inverted Pendulum

The idea of the sliding mode control is to use a discontinuous control through a variable structure system. A hypersurface

$$s(\varepsilon, \dot{\varepsilon}) = C^T x \quad (9)$$

defines the discontinuity manifold [11]. If we consider the approximation of the system (2) using Eq (5), the control is

$$F = \frac{1}{\tilde{g}} \left( -pS - \tilde{f} + \ddot{y}_d - \dot{\xi} \right) - \frac{K}{\tilde{g}} \text{sign}(s(t)). \quad (10)$$

$\xi$  is a linear control which ensures for a known system the condition :

$$\dot{s} = -ps,$$

where  $p$  is a chosen real positive number.

Calling  $\Delta f = f - \tilde{f}$  and  $\Delta g = g - \tilde{g}$ , it is

shown that the condition  $\dot{s}s < 0$ , which ensures the convergence of  $s$  to zero, is satisfied if the gain  $K$  is chosen big enough:

$$K \geq \left| \frac{\Delta f + \frac{\Delta g}{\tilde{g}} (-ps - \tilde{f} + \ddot{y}_d - \dot{\xi})}{1 + \frac{\Delta g}{\tilde{g}}} \right| \quad (11)$$

To avoid high frequency oscillations close to the sliding surface  $s(t)$ , it is proposed in [11] to replace the function  $\text{sign}(\cdot)$  in Eq (10) by the function saturation defined by:

$$\left\{ \text{sat}(x) = \frac{x}{\alpha}, \left| \frac{x}{\alpha} \right| < 1, \quad \text{sat}(x) = \text{sign}\left(\frac{x}{\alpha}\right), \left| \frac{x}{\alpha} \right| > 1. \right. \quad (12)$$

The value of  $\alpha$  is so chosen as to realise a trade-off between the robustness of the control and the damping of the chattering. In the simulations, the value of the parameters is

$$s(\varepsilon, \dot{\varepsilon}) = 5\varepsilon + \dot{\varepsilon}, p = 5, \alpha = 0.01, K = 4.$$

## 4. Piece-wise Affine Fuzzy Control

The membership functions  $\mu_j(\varepsilon)$  to the predicate if  $\varepsilon$  is  $A_j$  of the fuzzy controller defined in Section 2.A, are piece-wise affine functions of  $\varepsilon$ . It becomes obvious that the membership function  $\mu_{j,k}(\varepsilon, \dot{\varepsilon}) = \min(\mu_j(\varepsilon), \mu_k(\dot{\varepsilon}))$  to the predicate if  $\varepsilon$  is  $A_j$  AND  $\dot{\varepsilon}$  is  $B_k$  is a piece-wise linear function of vector  $x$ . Thus there exists a number  $M$  of regions  $\mathcal{R}_p$  with  $p$

$=1,2,\dots,M$ , in the plane  $\{\varepsilon, \dot{\varepsilon}\}$  where these membership functions are affine functions of  $x$ .

Yet, the centroid method defined in (8) yields a non-linear controller with constant parameters in each region  $\mathcal{R}_p$ .

Let us suppose that  $N$  rules are active in each region  $\mathcal{R}_p$  considered, and rule  $R_{j,k}$  reads:

$R_{j,k}$ : if  $\varepsilon$  is  $A_j$  AND  $\dot{\varepsilon}$  is  $B_k$  then  $F$  is  $F_{j,k}$ .

In order to remove the non-linearities due to the defuzzification stage, the following defuzzification operator is introduced:

$$F = \frac{1}{N} \sum_{j,k} (1 + \mu_{j,k} - \mu_{h(j,k)}) F_{j,k} \quad (13)$$

where  $\mu_{j,k}$  corresponds to the membership function for the rule  $R_{j,k}$  and the function  $h$  is defined as follows:

$$h: \{1..N\} \rightarrow \{1..N\}$$

$h(l) \neq l$ , and  $h$  is a bijection from  $\{1..N\}$  into  $\{1..N\}$ .

The belief in rule  $R_{j,k}$  is thus altered by the belief in another rule  $R_{h(j,k)}$ : in the inverted pendulum example, rules  $R_{1,1}$ ,  $R_{2,2}$ ,  $R_{2,1}$  and  $R_{1,2}$  are active at the same time in the quadrant  $\varepsilon < 0$  and  $\dot{\varepsilon} > 0$ . For all the regions  $\mathcal{R}_p$  included in this quadrant, we compensate the membership function of the cell of the look-up table which is far from the origin (that is to say rule  $R_{1,1}$ ) with the nearest (rule  $R_{2,2}$ ) and the other two active rules by each other (rule  $R_{2,1}$  and rule  $R_{1,2}$ ). The operation is repeated for the other quadrants as well.

In every region  $\mathcal{R}_p$  where the membership functions are piece-wise linear, the resulting fuzzy control also becomes piece-wise linear and can thus be written:

$$F(\varepsilon, \dot{\varepsilon}) = \alpha_p x + V_p,$$

for every region  $\mathcal{R}_p$ ,  $p=1 \dots M$ . (14)

$N$  is the number of active rules in the region  $\mathcal{R}_p$ . Since the  $\mu_{j,k}$  are piece-wise affine functions,  $V_p$  and  $\alpha_p$  are linear functions of the output membership functions  $F_{j,k}$ . As a result, there exist some parameters  $\beta_{j,k,p}$  such that:

$$V_p = \sum_{j,k} \beta_{j,k,p} F_{j,k} \quad (15)$$

The control defined above has a variable structure because the parameter value is region-wise. A simple way to obtain a stable controller is to tune the parameters ( $F_{j,k}$ ) so that a switching line separates the phase plane into one region where the control is positive, whereas in the other it is negative. In this case, it is possible to assimilate the controller to a Variable Structure Controller with Boundary Layer [9] as described in [7] [8]. The difference with the last scheme is that we may obtain a linear switching surface using the formalism of fuzzy control. We may thus choose the membership functions such that there exists a common frontier to some of the regions  $\mathcal{R}_p$ ,  $s(x) = C^T x$  where  $C$  is a constant matrix; in the regions  $\mathcal{R}_p$  where the surface  $C^T x$  is always positive (respectively negative), a constant strictly negative (respectively positive) term will be introduced using the term  $V_p$  in equation (7); if  $s(x)$  is both negative and positive in some region  $\mathcal{R}_p$ , then the term  $V_p$  in Eq (7) will be cancelled by tuning the control parameters  $F_{j,k}$ , the control being then linear in the regions considered. Considering the control of the linearised system (6), the algorithm will thus be:

$$\dot{x} = Ax + B \left( \alpha_p^T x + V_p \right), \quad (16)$$

where we choose  $V_p > 0$  if  $C^T x \leq 0$  for every  $x \in \mathcal{R}_p$ ,  $V_p < 0$  if  $C^T x \geq 0$  for every  $x \in \mathcal{R}_p$ , else  $V_p = 0$ .

The last condition can be rewritten, using Eq (7),

$$V_p = \sum_{j,k} \beta_{j,k,p} F_{j,k} = 0 \quad (17)$$

The control  $F = \alpha_p^T x + V_p$  may be discontinuous with respect to the surface  $s(x) = C^T x = 0$ . Eq (16) can be rewritten as:

$$\dot{x} = (A + B \alpha_p^T) x - B \left| V_p \right| \text{sign}(C^T x). \quad (18)$$

Let us suppose that  $B^T C > 0$ ; a sufficient condition for the stability of the system (19) is

that, for every region  $\mathcal{R}_p$ , the matrix  $Q_p = \left( (A+B\alpha_p^T)^T CC^T + CC^T (A+B\alpha_p^T) \right) < 0$ . (19)

The proof is straightforward using the candidate Lyapunov function  $V = s^T s$ . Derivating  $V$ , we obtain:

$$\begin{aligned} \dot{V} &= \dot{x}^T CC^T x + x^T CC^T \dot{x}, \\ \dot{V} &= x^T (A+B\alpha_p^T)^T CC^T x - B^T |V_p| \text{sign}(C^T x) CC^T x + x^T CC^T (A+B\alpha_p^T) \dot{x} - x^T CC^T B |V_p| \text{sign}(C^T x), \\ \dot{V} &= x^T \left( (A+B\alpha_p^T)^T CC^T + CC^T (A+B\alpha_p^T) \right) x - 2(B^T C) |V_p| \text{sign}(C^T x) (C^T x) \end{aligned}$$

Obviously,  $\dot{V} < 0$  if condition (19) is met.

### C. Application to the Inverted Pendulum

Let us consider the quadrant called  $Q_1$  where  $-a < \varepsilon < 0$  and  $0 < \dot{\varepsilon} < b$ ; only rules  $R_{1,1}$ ,  $R_{2,1}$ ,  $R_{1,2}$  and  $R_{2,2}$  are active. The membership functions for these rules are respectively,

$$\begin{aligned} \mu_{R_{1,1}} &= \min\left(-\frac{\varepsilon}{a}, \frac{\dot{\varepsilon}}{b}\right), \quad \mu_{R_{2,1}} = \min\left(1 + \frac{\varepsilon}{a}, \frac{\dot{\varepsilon}}{b}\right), \\ \mu_{R_{1,2}} &= \min\left(1 + \frac{\varepsilon}{b}, 1 - \frac{\dot{\varepsilon}}{b}\right), \quad \mu_{R_{2,2}} = \min\left(-\frac{\varepsilon}{a}, 1 - \frac{\dot{\varepsilon}}{b}\right) \end{aligned}$$

Thus we have 4 regions  $\mathcal{R}_1, \mathcal{R}_2, \mathcal{R}_3, \mathcal{R}_4$ , for which the membership functions are all linear with respect to  $\varepsilon$  and  $\dot{\varepsilon}$  and have constant parameters (Figure 3).

$$\begin{aligned} \mathcal{R}_1: & -\frac{\varepsilon}{a} > \frac{\dot{\varepsilon}}{b} \text{ and } 1 + \frac{\varepsilon}{a} < \frac{\dot{\varepsilon}}{b}, & \mathcal{R}_2: & -\frac{\varepsilon}{a} < \frac{\dot{\varepsilon}}{b} \text{ and } 1 + \frac{\varepsilon}{a} < \frac{\dot{\varepsilon}}{b}, \\ \mathcal{R}_3: & -\frac{\varepsilon}{a} < \frac{\dot{\varepsilon}}{b} \text{ and } 1 + \frac{\varepsilon}{a} > \frac{\dot{\varepsilon}}{b}, & \mathcal{R}_4: & -\frac{\varepsilon}{a} > \frac{\dot{\varepsilon}}{b} \text{ and } 1 + \frac{\varepsilon}{a} > \frac{\dot{\varepsilon}}{b}. \end{aligned}$$

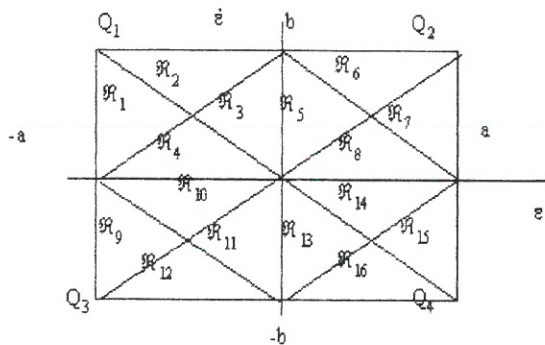


Figure 3. Regions with Constant Parameters Membership Functions

For example, for region  $\mathcal{R}_1$ , we have:

$$\mu_{R_{1,1}} = \frac{\dot{\varepsilon}}{b}, \quad \mu_{R_{1,2}} = 1 + \frac{\varepsilon}{a}, \quad \mu_{R_{2,2}} = 1 + \frac{\varepsilon}{a}, \quad \mu_{R_{2,1}} = 1 - \frac{\dot{\varepsilon}}{b}.$$

This operation can be repeated for the

other three quadrants  $0 < \varepsilon < a$  and  $-b < \dot{\varepsilon} < 0$ ,  $-a < \varepsilon < 0$  and  $0 < \dot{\varepsilon} < b$ ,  $0 < \varepsilon < a$  and  $0 < \dot{\varepsilon} < b$  which finally makes 16 regions where

the  $\mu_{R_{i,j}}$  has a constant expression (see Figure 3).

The resulting control in the quadrant  $Q_1$  ( $-a < \varepsilon < 0$  and  $0 < \dot{\varepsilon} < b$ ) is then:

$$\begin{aligned} F &= \frac{1}{2} (1 + \mu_{R_{1,1}} - \mu_{R_{2,1}}) F_{1,1} + \frac{1}{2} (1 + \mu_{R_{1,2}} - \mu_{R_{2,1}}) \\ &F_{1,2} + \frac{1}{2} (1 + \mu_{R_{2,1}} - \mu_{R_{1,1}}) F_{2,1} + \frac{1}{2} (1 + \mu_{R_{2,2}} - \mu_{R_{1,1}}) F_{2,2}. \end{aligned}$$

Replacing the expression of the  $\mu_{R_{i,j}}$  in the region  $\mathcal{R}_1$ , we have

$$\mathcal{R}_1: F = \frac{1}{2} \left( \begin{aligned} &\left(\frac{\dot{\varepsilon}}{b} - \frac{\varepsilon}{a}\right) F_{1,1} + \left(1 + \frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a}\right) F_{1,2} + \left(1 - \frac{\varepsilon}{a} - \frac{\dot{\varepsilon}}{b}\right) \\ &F_{2,1} + \left(2 + \frac{\varepsilon}{a} - \frac{\dot{\varepsilon}}{b}\right) F_{2,2} \end{aligned} \right) \quad (20)$$

Due to the symmetry of the operator and of the membership function, the expression for control  $F$  is the same in regions  $\mathcal{R}_2, \mathcal{R}_3$ , and  $\mathcal{R}_4$ , so that the above formula holds for the whole quadrant  $-a < \varepsilon < 0$  and  $0 < \dot{\varepsilon} < b$ .

For the other quadrants, using the same method, we get:

$$Q_2: \quad 0 < \varepsilon < a \text{ and } 0 < \dot{\varepsilon} < b, \\ F = \frac{1}{2} \left( \begin{aligned} &\left(\frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a}\right) F_{1,3} + \left(1 - \frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a}\right) F_{2,3} + \\ &\left(1 - \frac{\varepsilon}{a} + \frac{\dot{\varepsilon}}{b}\right) F_{1,2} + \left(2 - \frac{\varepsilon}{a} - \frac{\dot{\varepsilon}}{b}\right) F_{2,2} \end{aligned} \right),$$

$$Q_3: \quad -a < \varepsilon < 0 \text{ and } -b < \dot{\varepsilon} < 0, \\ F = \frac{1}{2} \left( \begin{aligned} &\left(-\frac{\dot{\varepsilon}}{b} - \frac{\varepsilon}{a}\right) F_{3,1} + \left(1 - \frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a}\right) F_{2,1} + \\ &\left(1 + \frac{\varepsilon}{a} - \frac{\dot{\varepsilon}}{b}\right) F_{3,2} + \left(2 + \frac{\varepsilon}{a} + \frac{\dot{\varepsilon}}{b}\right) F_{2,2} \end{aligned} \right),$$

$Q_4: 0 < \varepsilon < a$  and  $-b < \dot{\varepsilon} < 0$ ,

$$F = \frac{1}{2} \left( \begin{array}{l} \left( -\frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a} \right) F_{3,3} + \left( 1 + \frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a} \right) F_{2,3} + \\ \left( 1 - \frac{\varepsilon}{a} - \frac{\dot{\varepsilon}}{b} \right) F_{3,2} + \left( 2 - \frac{\varepsilon}{a} + \frac{\dot{\varepsilon}}{b} \right) F_{2,2} \end{array} \right)$$

The controller is region-wise linear in the four quadrants, the symmetry of the membership functions reduces the number of the regions to be considered.

#### D. Stability Analysis and Synthesis

The control (20) is thus piece-wise affine with constant parameters in each of the four quadrants  $Q_1, Q_2, Q_3$  and  $Q_4$ . A common frontier to some of the regions  $\mathcal{R}_i$  is the surface:

$$s(x) = \frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a} = 0, \quad (21)$$

The control will not be discontinuous along this surface since it belongs to quadrants  $Q_1$  and  $Q_4$  where the control is continuous. However, we can choose the controls to be discontinuous around these quadrants, that is to say in  $Q_2$  and  $Q_3$ . We can thus choose the control to be linear in quadrants  $Q_1$  and  $Q_4$  ( $V_i = 0$ ) and have a discontinuous term in  $Q_2$  and  $Q_3$  so that, using Eq (16),  $V_p \neq 0$  and  $V_p \cdot s(x) \geq 0$ . The application of the method yields, as an application of formula (17):

$$\begin{aligned} Q_1 : F_{1,2} + F_{2,1} + 2F_{2,2} &= 0, \\ Q_4 : F_{3,2} + F_{2,3} + 2F_{2,2} &= 0 \end{aligned} \quad (22)$$

and in the discontinuity parts:

$$\begin{aligned} Q_2 : V_2 = F_{1,2} + F_{2,3} + 2F_{2,2} &< 0, \\ Q_3 : V_3 = F_{3,2} + F_{2,1} + 2F_{2,2} &> 0 \end{aligned} \quad (23)$$

Substituting Eq (22) in Eq (23) yields:

$$\begin{aligned} Q_2 : V_2 = F_{2,3} - F_{2,1} &< 0, \\ Q_3 : V_3 = F_{2,1} - F_{2,3} &> 0 \end{aligned} \quad (24)$$

To satisfy equation (24), it suffices to take  $F_{2,1} \neq F_{2,3}$ . The expression of the controller is this time:

$$\left\{ \begin{array}{l} Q_1 : u = \frac{1}{2} \left( \left( \frac{\dot{\varepsilon}}{b} - \frac{\varepsilon}{a} \right) F_{1,1} + F_{2,1} + \left( -\frac{3\dot{\varepsilon}}{b} - \frac{\varepsilon}{a} \right) F_{2,2} \right) \\ Q_2 : u = \frac{1}{2} \left( \left( \frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a} \right) F_{1,3} + \left( -\frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a} \right) F_{4,1} + \right. \\ \left. \left( \frac{\varepsilon}{a} - \frac{\dot{\varepsilon}}{b} \right) F_{2,3} + F_{2,3} - F_{2,1} \right) \\ Q_3 : u = \frac{1}{2} \left( \left( -\frac{\dot{\varepsilon}}{b} - \frac{\varepsilon}{a} \right) F_{3,1} + \left( \frac{\dot{\varepsilon}}{b} - \frac{\varepsilon}{a} \right) F_{2,1} + \right. \\ \left. \left( -\frac{\varepsilon}{a} + \frac{\dot{\varepsilon}}{b} \right) F_{2,3} + F_{2,1} - F_{2,3} \right) \\ Q_4 : u = \frac{1}{2} \left( \left( -\frac{\dot{\varepsilon}}{b} + \frac{\varepsilon}{a} \right) F_{3,3} + \left( \frac{2\dot{\varepsilon}}{b} + \frac{2\varepsilon}{a} \right) F_{2,3} + \right. \\ \left. \left( \frac{3\dot{\varepsilon}}{b} + \frac{\varepsilon}{a} \right) F_{2,2} \right) \end{array} \right. \quad (25)$$

For the inverted pendulum stabilisation, we take  $a = b = 0.02$ . The surface  $s(x)$  will thus be  $C^T x = \dot{\varepsilon} + \varepsilon = 0$ .

With the choice of  $F_{1,1} = 1, F_{1,3} = 18, F_{2,1} = -0.55, F_{2,3} = 0.85, F_{3,1} = -10, F_{3,3} = 0.1, F_{2,2} = -0.1$  we obtain the following matrices

$$Q_p = \left( \left( A + B\alpha_p^T \right)^T C C^T + C C^T \left( A + B\alpha_p^T \right) \right)$$

corresponding to the quadrants  $Q_p$ :

$$\begin{aligned} Q_1 &= \begin{pmatrix} -1.85 & -0.96 \\ -0.96 & -0.07 \end{pmatrix}, & Q_2 &= \begin{pmatrix} -318 & -306 \\ -306 & -295 \end{pmatrix}, \\ Q_3 &= \begin{pmatrix} -185 & -168 \\ -168 & -152 \end{pmatrix}, & Q_4 &= \begin{pmatrix} -9.56 & -4.8 \\ -4.8 & -0.07 \end{pmatrix}. \end{aligned}$$

It can be checked that  $Q_p < 0$  for every  $j$  and, following condition (19), the overall system is stable.

To avoid chattering due to the discontinuity term, we use the  $\text{sat}(\cdot)$  function defined in (12) with  $\alpha = 0.03$ .

As for the VSC controller defined in Section 3. A, it is to hope that the modelling error will be small enough so that the convergence of the overall system should not be affected.

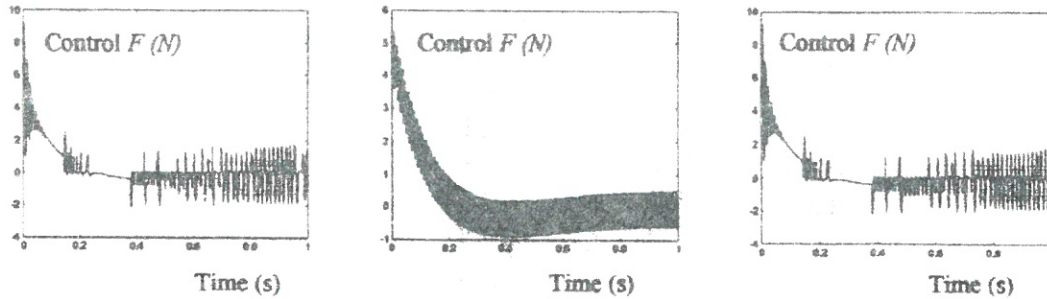


Figure 4. (a) Fuzzy Control, (b) Variable Structure Control, (c) Piece-wise Affine Fuzzy Control

## 5. Numerical Results

The structure of the piece-wise affine controller is now clear: a fast tracking is allowed by the "discontinuity" terms, which are region-wise, and the regulation is ensured nearby the surface by linear terms. The Sliding Mode Controller described in [9] has a better performance but it might rather difficult to design as the equations of the system should be known. The controller also resembles the one which was described in [7] or [8] which can be

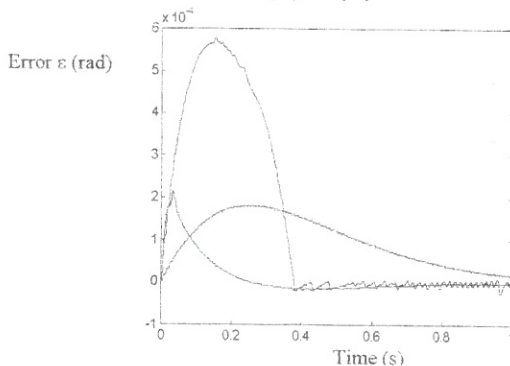


Figure 5. Control Errors  
Fuzzy Control, Variable Structure  
Control,  
Piece-wise Affine Fuzzy Control

seen as an extension of the previously cited controller, but it has the advantage of being piece-wise linear and displaying a linear switching surface which is a great simplification. A major advantage over the classical fuzzy controller lies in that Eqs (17) and (19) provide a framework for a systematic controller design. These equations provide some inequalities which restrict the admissible parameter domain and give stability or performance conditions. Parameter tuning can thus be immediately compared to the trial-and-error method used for fuzzy control

## 6. Conclusion

A linear defuzzification method has allowed to turn a fuzzy controller into a region-wise affine state space controller; the control output has a region-wise constant term and a region-wise term which is a linear function of the controller inputs. A systematic parameter tuning ensures the stability of the controller, and can be of a real help for control synthesis. It was shown that this controller could be seen as a particular case of variable structure control with boundary layer.

The performance and interests of the control was shown through the control of an inverted pendulum and compared to variable structure control and fuzzy control. A simple defuzzification configuration was tested, and future work was suggested for other strategies.

## REFERENCES

1. LEE, C. C., **Fuzzy Logic in Control Systems: Fuzzy Logic Controller I-II**, IEEE TRANSACTIONS ON SYSTEMS, MAN AND CYBERNETICS, 20, 1990, pp.404-418.
2. TANAKA, K. and SUGENO, M., **Stability Analysis and Design of Fuzzy Control Systems**, FUZZY SETS SYSTEMS, Vol. 45, 1992, pp. 135-156.
3. YING, H., **Practical Design of Nonlinear Fuzzy Controllers With Stability Analysis for Regulating Processes With Unknown Mathematical Models**, AUTOMATICA, Vol. 30, 1994, pp. 1185-1195.
4. DE NEYER, M. and GOREZ, R., **Comments on Practical Design of Nonlinear Fuzzy**

- Controllers With Stability Analysis for Regulating Processes With Unknown Mathematical Models**, AUTOMATICA, Vol. 32, 1996, pp. 1613-1614.
5. DIEULOT, J.-Y., HAJRI, S. and BORNE, P., **Fuzzy Control Synthesis With the Help of Fuzzy Algorithms**, STUDIES IN INFORMATICS AND CONTROL, Vol. 5, No. 1, March 1996, PP 5-13.
  6. GLOVER, J. S. and MUNIGHAN, J., **Designing Fuzzy Controllers From A Variable Structure Standpoint**, IEEE TRANSACTIONS ON FUZZY SYSTEMS, Vol. 5, 1997, pp. 138-144.
  7. PALM, R., **Robust Control By Sliding Mode**, AUTOMATICA, 30, 1994, pp.1429-1437.
  8. PALM, R., **Sliding Mode Fuzzy Control** , IEEE International Conference on Fuzzy Systems, San Diego, CA, USA, 1992, pp. 519-526.
  9. SLOTINE , J. J. E., **The Robust Control of Robot Manipulators**, THE INTERNATIONAL JOURNAL OF ROBOTICS RESEARCH, Vol. 4, 1985, pp. 49-64.
  10. KOSKO, B., **Neural Networks and Fuzzy Systems**, PRENTICE -HALL, 1992.
  11. SLOTINE, J. J. E. and SASTRY, S. S., **Tracking Control Problem of Non-linear Systems Using Sliding Surfaces With Application to Robots Manipulators**, INTERNATIONAL JOURNAL OF CONTROL, Vol. 38, 1985, pp. 110-132.