# FUZZY CONTROL PI VS. FUZZY PI-CONTROL

**Olaf Wolkenhauer** 

**Control Systems Centre** 



o.wolkenhauer@umist.ac.uk

www.csc.umist.ac.uk/people/wolkenhauer.htm

### Contents

44

◀

1	Learning Objectives								
<b>2</b>	Feedback Control								
3	Fuzzy PI-Controller	6							
4	Example: First-Order System with Dead-Time4.1Fuzzy Rule-Base4.2Centre Average Defuzzification4.3Control Surface: Zadeh Logic4.4Analysis4.5Summary4.6First-Order Delaved Process	<b>15</b> 17 18 19 20 29 30							
5	<ul> <li>Example: Coupled Tanks</li> <li>5.1 Design of a fluid level proportional controller</li> <li>5.2 Design of a proportional plus integral controller</li> </ul>	<b>33</b> 34 35							

Back

View

 $\mathbf{2}$ 

53	Design of a	fuzzy-PI-controller													38
0.0	Design Of a	Tuzzy-1 1-controller.	•	•	• •	•	•	•	•	•	•	•	•	•	00



## 1. Learning Objectives

- □ Fuzzy rule-based systems can also be used to devise control laws.
- □ Fuzzy control can be particular useful if no linear parametric model of the process under control is available.
- □ Fuzzy control is not 'model-free' as a good understanding of the process dynamics may be required.
- □ Fuzzy control lacks of design methodologies.
- $\Box$  Fuzzy controllers are easy to understand and simple to implement.



## 2. Feedback Control

**Feedback:** When we desire a system to follow a given pattern the difference between this pattern and the actual behaviour is used as a new input to cause the part regulated to change in such a way as to bring its behaviour closer to that given by the pattern.





### 3. Fuzzy PI-Controller





▷ Conventional PI-Control:

$$\frac{\mathrm{d}u(t)}{\mathrm{d}t} = K_p \ \frac{\mathrm{d}e(t)}{\mathrm{d}t} + K_i \ e(t) \tag{1}$$
 where  $e(t) = s(t) - y(t).$ 

- ▷ To obtain a control action the term du(t)/dt is integrated.
- ✗ A fuzzy-PI-controller is developed analogously :

$$deriv'(k) = K_p \cdot rate'(k) + K_i \cdot error'(k)$$
(2)

▷ error, rate, deriv are fuzzy (or linguistic) variables partioning the underlying spaces by piecewise linear (triangular) fuzzy sets as shown in figure 1.





Figure 1: Fuzzy sets for the variables error, rate and the output.



- $\triangleright$  Scale (2) to a range from -1 to +1.
- $\triangleright$  Use scaling factors  $G_u$ ,  $G_r$  and  $G_e$ , where

 $deriv' = G_u \cdot deriv, \quad G_r \cdot rate' = rate, \quad G_e \cdot error' = error$ 

 $\triangleright$  Substituting these into (2)

$$deriv(k) = \frac{K_p}{G_u G_r} rate(k) + \frac{K_i}{G_u G_e} error(k)$$
(3)

- ▷ The constants  $K_p$  ( $G_uG_r$ ) and  $K_i/(G_uG_e)$  are assumed equal to 0.5 to make *deriv* fall into the interval [-1, 1].
- ★ The fuzzy controller is then equivalent to a conventional PIcontroller with proportional gain  $K_p = 0.5 \cdot G_u \cdot G_r$  and integral gain  $K_i = 0.5 \cdot G_u \cdot G_e$ .
- ▷ Note: there are infinitely many combinations of  $G_e$ ,  $G_r$ , and  $G_u$  to hold true for these expressions.



The complete rule-base:

- $R_1$ : IF *error* is 'negative' AND *rate* is 'negative', THEN *deriv* is 'negative large'
- $R_2$ : IF *error* is 'negative' AND *rate* is 'zero', OR *error* is 'zero' AND *rate* is 'negative', THEN *deriv* is 'negative small'
- $\begin{array}{ll} R_3: & \text{IF $error$ is `negative' AND $rate$ is `positive',} \\ & \text{OR $error$ is `zero' AND $rate$ is `zero',} \\ & \text{OR $error$ is `positive' AND $rate$ is `negative',} \\ & \text{THEN $deriv$ is `zero'$} \end{array}$
- $R_4$ : IF *error* is 'zero' AND *rate* is 'positive', OR *error* is 'positive' AND *rate* is 'zero', THEN *deriv* is 'positive small'
- $R_5$ : IF *error* is 'positive' AND *rate* is 'positive', THEN *deriv* is 'positive large'



- $\triangleright$  'negative', 'zero', 'positive', etc. are fuzzy sets.
- ▷ The logical connectives 'AND' and 'OR' are are *t* and *t*-conorms.
- $\triangleright$  *firing level* of the *i*<sup>th</sup> rule, denoted  $\mu_{\text{DERIV}_i}(deriv)$ .
- ▷ Assuming  $n_s$  fuzzy sets for 'error', 'error rate of change' we require  $2(n_s 1)$  fuzzy sets (and rules) for the output *deriv*.

		Ν	$\mathbf{Z}$	Р
	Р	Ζ	PS	PL
rate	Ζ	NS	Ζ	PS
	Ν	NL	NS	Ζ

#### error



- ▷ The principal values for which  $\mu_{\text{DERIV}_i}(deriv) = 1$ , are equally spaced, but at half the interval of the antecedent fuzzy sets.
- ▷ With three fuzzy sets on the input spaces. The principal values of the  $i^{\text{th}}$  member of the fuzzy partition DERIV<sub>i</sub> are given by

$$-1 + (i-1)/(n_s - 1)$$

▷ Linear defuzzification strategy :

$$deriv(k) = \sum_{i=1}^{2n_s - 1} \mu_{\text{DERIV}_i}(deriv) \cdot \left( -1 + \frac{(i-1)}{n_s - 1} \right) .$$
(4)

▷ This value is integrated and scaled to obtain the control action required to drive the plant.



-0.5

-1 -1

Zadeh-logic:





-0.8 -0.6 -0.4 -0.2

0.2

0.4 0.6 0.8

*Mixed-logic* : Rule 1 and 5: Zadeh-logic Rule 2 and 4: Lukasiewicz-logic

> Conjunction:  $T(\mu_A(\cdot), \mu_B(\cdot)) = \max(0, (\mu_A(\cdot) + \mu_B(\cdot)) - 1)$ Disjunction:  $S(\mu_A(\cdot), \mu_B(\cdot)) = \min(1, \mu_A(\cdot) + \mu_B(\cdot))$





### 4. Example: First-Order System with Dead-Time

 $\triangleright$  Replace linear defuzzification by non-linear strategy [5].

▷ Notation : (sampling period equals one)

$$error'(k) = s(k) - y(k)$$

$$error(k) = G_e \cdot error'(k)$$

$$rate'(k) = error'(k) - error'(k-1)$$

$$rate(k) = G_r \cdot rate'(k)$$

$$deriv'(k) = G_u \cdot deriv(k)$$

$$u(k) = u(k-1) + deriv'(k) .$$

 $\triangleright$  Input and output fuzzy sets.



#### Input (error, rate) and output fuzzy sets:





### 4.1. Fuzzy Rule-Base

There are three fuzzy control rules composed out of four :

- $R_1$ : IF *error* is 'negative' AND *rate* is 'negative', THEN *deriv* is 'negative'
- $R_2$ : IF *error* is 'negative' AND *rate* is 'positive', THEN *deriv* is 'zero'
- $R_3$ : IF *error* is 'positive' AND *rate* is 'negative', THEN *deriv* is 'zero'
- $R_4$ : IF *error* is 'positive' AND *rate* is 'positive', THEN *deriv* is 'positive'



#### 4.2. Centre Average Defuzzification

Instead of

$$deriv(k) = \sum_{i=1}^{2n_s - 1} \mu_{\text{DERIV}_i}(deriv) \cdot \left( -1 + \frac{(i-1)}{n_s - 1} \right) .$$
(4)

normalise the membership degrees to one :

$$deriv(k) = \frac{\sum_{i=1}^{2n_s-1} \mu_{DERIV_i}(deriv) \cdot \left(-1 + \frac{i-1}{n_s-1}\right)}{\sum_{i=1}^{2n_s-1} \mu_{DERIV_i}(deriv)}$$
(5)

... called Center Average Defuzzification.



#### 4.3. Control Surface: Zadeh Logic



✗ Compare with linear defuzzification!



#### 4.4. Analysis

The membership functions associated with 'error' and 'rate of change' of the error :

- - -

$$\mu_{error is pos}(e(k)) = \frac{error(k) + 1}{2} = \frac{G_e \cdot error'(k) + 1}{2}$$
(6)  

$$\mu_{error is neg}(e(k)) = \frac{-error(k) + 1}{2} = \frac{-G_e \cdot error'(k) + 1}{2}$$
(7)  

$$\mu_{rate is pos}(r(k)) = \frac{rate(k) + 1}{2} = \frac{G_r \cdot rate'(k) + 1}{2}$$
(8)  

$$\mu_{rate is neg}(r(k)) = \frac{-rate(k) + 1}{2} = \frac{-G_r \cdot rate'(k) + 1}{2}$$
(9)



Partition of the phase-plane into sectors :

Sector	$R_1$	$R_2$	$R_3$	$R_4$
1	error is 'neg.'	rate is 'neg.'	error is 'neg.'	rate is 'pos.'
2	error is 'neg.'	rate is 'neg.'	error is 'neg.'	rate is 'pos.'
3	rate is 'neg.'	rate is 'neg.'	error is 'neg.'	<i>error</i> is 'pos.'
4	rate is 'neg.'	rate is 'neg.'	error is 'neg.'	<i>error</i> is 'pos.'
5	rate is 'neg.'	<i>error</i> is 'pos.'	rate is 'pos.'	<i>error</i> is 'pos.'
6	rate is 'neg.'	<i>error</i> is 'pos.'	rate is 'pos.'	<i>error</i> is 'pos.'
7	error is 'neg.'	<i>error</i> is 'pos.'	rate is 'pos.'	rate is 'pos.'
8	error is 'neg.'	<i>error</i> is 'pos.'	rate is 'pos.'	rate is 'pos.'



Partioning of the phase-plane for a fuzzy PI-controller :



From equations (6)-(9) and (5) we obtain the following equations for the output :

#### Sector 1 and 2 :

$$deriv(k) = \frac{-\mu_{error \ is \ neg}(e(k)) + \mu_{rate \ is \ pos}(r(k))}{\mu_{er. \ is \ neg}(e(k)) + \mu_{rate \ is \ neg}(r(k)) + \mu_{rate \ is \ pos}(r(k))}$$
$$= \frac{G_e \cdot error'(k) + G_r \cdot rate'(k)}{3 - G_e \cdot error'(k)} \tag{10}$$

#### Sector 3 and 4 :

$$deriv(k) = \frac{-\mu_{rate \ is \ neg}(r(k)) + \mu_{err \ is \ pos}(e(k))}{\mu_{rate \ is \ neg}(r(k)) + \mu_{er. \ is \ neg}(e(k)) + \mu_{er. \ is \ pos}(e(k))}$$
$$= \frac{G_r \cdot rate'(k) + G_e \cdot error'(k)}{3 - G_r \cdot rate'(k)} \tag{11}$$



#### Sector 5 and 6 :

$$deriv(k) = \frac{-\mu_{rate \ is \ neg}(r(k)) + \mu_{err \ is \ pos}(e(k))}{\mu_{rate \ is \ neg}(r(k)) + \mu_{rate \ is \ pos}(r(k)) + \mu_{er. \ is \ pos}(e(k))}$$
$$= \frac{G_r \cdot rate'(k) + G_e \cdot error'(k)}{3 + G_r \cdot error'(k)}$$
(12)

Sector 7 and 8 :

$$deriv(k) = \frac{-\mu_{error \ is \ neg}(e(k)) + \mu_{rate \ is \ pos}(r(k))}{\mu_{er. \ is \ neg}(e(k)) + \mu_{er. \ is \ pos}(e(k)) + \mu_{rate \ is \ pos}(r(k))}$$
$$= \frac{G_e \cdot error'(k) + G_r \cdot rate'(k)}{3 + G_r \cdot rate'(k)} \tag{13}$$



If  $G_r |rate'(k)| \le G_e |error'(k)| \le 1$ , we then have  $deriv(k) = \frac{G_e \cdot error'(k) + G_r \cdot rate'(k)}{3 - G_e \cdot |error'(k)|}$ (14)

and if  $G_e|error'(k)| \le G_r|rate'(k)| \le 1$ ,

$$deriv(k) = \frac{G_e \cdot error'(k) + G_r \cdot rate'(k)}{3 - G_r \cdot |rate'(k)|} .$$
(15)



▷ As we have seen in the previous section, the fuzzy PI-controller with linear defuzzification and mixed logic is equivalent to a nonfuzzy PI-controller with proportional gain  $K_p = 0.5 \cdot G_u \cdot G_r$ and integral gain  $K_i = 0.5 \cdot G_u \cdot G_e$ :

$$deriv'(k) = K_p \cdot rate'(k) + K_i \cdot error'(k) .$$
 (16)

▷ Comparing (16) with equations (14) and (15), we notice that the fuzzy PI-controller with nonlinear defuzzification and Zadehlogic for rule evaluation is equivalent to a linear PI-controller with changing gains  $K_p$  and  $K_i$ :

$$K_p = \frac{G_r \cdot G_u}{3 - G_e |error'(k)|} \tag{17}$$

$$K_i = \frac{G_e \cdot G_u}{3 - G_e |error'(k)|} \tag{18}$$



when  $G_r |rate'(k)| \le G_e |error'(k)| \le 1$ , and  $K_p = \frac{G_r \cdot G_u}{3 - G_r |rate'(k)|}$   $K_i = \frac{G_e \cdot G_u}{3 - G_r |rate'(k)|}$ (19)
(20)

when  $G_e|error'(k)| \leq G_r|rate'(k)| \leq 1$ .

▷ If we define the static gains  $K_{p_s}$  and  $K_{i_s}$  as the proportional and integral gains when both *error'* and *rate'* are equal to zero, we have :

$$K_{p_s} = \frac{G_r \cdot G_u}{3} \tag{21}$$
$$K_{t-1} = \frac{G_e \cdot G_u}{3} \tag{22}$$

$$K_{i_s} = \frac{C_e - C_u}{3} \tag{22}$$



and find for the conventional PI-controller

$$deriv(k) = \frac{K_{p_s}}{G_u} \cdot rate'(k) + \frac{K_{i_s}}{G_u} \cdot error'(k)$$
$$= \frac{G_r \cdot rate'(k) + G_e \cdot error'(k)}{3} .$$
(23)

Back

View

28

▷ Comparing equality (23) with equations (14) and (15), the following inequalities are obtained :

$$\begin{aligned} \frac{1}{3 - G_e \cdot error'(k)} &\geq \frac{1}{3} \\ \text{when } G_r | rate'(k) | \leq G_e | error'(k) | \leq 1, \text{ and} \\ \frac{1}{3 - G_r \cdot rate'(k)} &\geq \frac{1}{3} \\ \text{when } G_r | error'(k) | \leq G_e | rate'(k) | \leq 1. \end{aligned}$$

### 4.5. Summary

- ▷ The (absolute value of the) incremental control action of the fuzzy PI-controller is equal or greater the (absolute value of the) incremental control action of the nonfuzzy PI-controller when  $G_e|error'(k)| \leq 1$  and  $G_r|rate'(k)| \leq 1$ .
- ▷ We can conclude that the larger (absolute values of) error (rate) values, the larger is the difference between the outputs of the two controllers.
- ▷ The nonlinearity of the fuzzy PI-controller can therefore be used to improve the control performance in comparison to a nonfuzzy and linear PI-controller.



#### 4.6. First-Order Delayed Process

- ▷ Comparison: Fuzzy PI-controller with Zadeh logic and nonlinear defuzzification vs linear PI-controller.
- ▷ Static proportional gain  $K_{p_s}$  and the static integral gain  $K_{i_s}$  of the fuzzy controller were set equal the proportional and integral gains  $K_p = 2.38$  and  $K_i = 4.43$  of the conventional PI-controller.
- ▷ The process plant is taken to be a first order system with time delay and transfer function ;

$$\frac{Y(s)}{U(s)} = \frac{1}{s+1} \cdot e^{-0.2s}$$



Step responses...



Phase plane and trajectory...



### 5. Example: Coupled Tanks

- $\triangleright$  The control input is the pump drive voltage.
- $\triangleright$  The sensed output is the water depth in tank 2.





#### 5.1. Design of a fluid level proportional controller

- ▷ Plot the root locus of the open loop transfer function  $G_v(z)$ , select the gain  $K_p$  that gives a closed loop damping factor of  $\vartheta = 0.7$ .
- $\triangleright$  Read off the c.l. natural frequency  $\omega_n$ .
- $\triangleright~$  The closed loop transfer function is :

$$H_v(z) = \frac{v_{d_2}(z)}{v_r(z)} = \frac{K_p \ G_v(z)}{1 + K_p \ G_v(z)}$$
(24)

where :

$$G_v(z) = \frac{z-1}{z} Z\left(\frac{G_v(s)}{s}\right)$$
(25)

▷ The steady state error can be calculated using :

$$e_{ss} = [v_r(z) - v_{d_2}(z)]_{z \to 1} = v_r(1) [1 - H_v(z)]_{z \to 1}$$
(26)

Where  $v_r(1)$  is the steady state reference input.



#### 5.2. Design of a proportional plus integral controller

- ▷ Set the integral action time constant to a reasonable value (in this case  $T_i = 50$ s).
- ▷ Plot the root locus of the open loop system in cascade with the compensator  $1 + \frac{1}{T_i} \frac{z}{z-1}$ , and select the gain k that gives a closed loop damping factor of  $\vartheta = 0.7$ .
- $\triangleright$  The proportional and integral gains  $K_p$ , and  $K_i$  can be computed by comparing coefficients of the compensator transfer functions:

$$K\left(1 + \frac{z}{(z-1)T_i}\right) = K_p + \frac{K_i z}{(z-1)}$$
 (27)

 $\triangleright$  The value of  $\omega_n$  can be read off the root locus plot.



Section 5: Example: Coupled Tanks

 $\triangleright$  Writing the open loop system in series with a proportional plus integral action compensator as  $G_c(s)$ , where :

$$G_c(s) = \frac{K\left(sT_i + 1\right)}{sT_i} \cdot \frac{\frac{g_p g_{d_2}}{K_2}}{T_1 T_2 s^2 + (T_1 + T_2) s + 1}$$
(28)

 $\triangleright$  The closed loop transfer function is :

$$H_{v}(s) = \frac{\frac{Kg_{p}g_{d_{2}}}{k_{2}} \cdot sT_{i} + 1}{T_{i}T_{1}T_{2}s^{3} + T_{i}\left(T_{1} + T_{2}\right)s^{2} + T_{i}\left(1 + \frac{Kg_{p}g_{d_{2}}}{K_{2}}\right)s + \frac{Kg_{p}g_{d_{2}}}{K_{2}}}$$
(29)

 $\triangleright$  Assuming :

$$T_1 > 0; \quad T_2 > 0; \quad g_p > 0; \quad g_{d_2} > 0; \quad k_2 > 0; \quad T_i > 0; \quad K > 0$$
(30)



Section 5: Example: Coupled Tanks

The closed loop system is stable for :

$$T_i \left( T_1 + T_2 \right) \left( 1 + \frac{Kg_p g_{d_2}}{K_2} \right) > T_1 T_2 \frac{Kg_p g_{d_2}}{K_2} \tag{31}$$

▷ Hence the closed loop system can become unstable for sufficiently large gain if :

$$T_i < \frac{T_1 T_2}{T_1 + T_2} \tag{32}$$

and the gain required to make the closed loop system unstable is :

$$K = \frac{K_2}{g_p g_{d_2}} \cdot \frac{T_i \left(T_1 + T_2\right)}{T_1 T_2 - T_i \left(T_1 + T_2\right)}$$
(33)

▷ The closed loop system under proportional plus integral control has three poles, and one zero. Using the design procedure outlined above, two of the poles will be complex conjugate (at  $\vartheta = 0.7$ ), the remaining pole and zero are negative real.



### 5.3. Design of a fuzzy-PI-controller.

- ▷ The constants  $\frac{K_p G_r}{G_u}$  and  $\frac{K_i G_e}{G_u}$  are set to make the output to fall within the interval [-1, +1].
- $\triangleright$  Choose fuzzy sets for the input and output as in figure 1.
- $\triangleright$  For example,  $G_e = 0.02$ ,  $G_r = 1$ , and  $G_u = 15$ .
- ▷ Fix the number of (triangular, fully overlapping) fuzzy sets partitioning the input spaces.
- ▷ On the basis of the phase-plane characteristic and/or trajectory decide upon the fuzzy logic employed. This will decide the overall gain structure of the phase-plane.
- $\triangleright$  Adjust input-output gains to have trajectories of the system to fall within the [-1, 1] range.
- ▷ Change positions of principal values for input fuzzy sets to finetune gain structure in the quadrants of the phase-plane.

Simulink block diagram :





**Step responses:** non-fuzzy PI-controller,  $K_i = 0.15$ ,  $K_p = 7.5$ , (dashed line) and its fuzzy equivalent :



## References

- Babuska, R.: Fuzzy Modelling for Control. Kluwer, 1998. See http://lcewww.et.tudelft.nl/.
- [2] Passino, K.M. and Yurkovich, S. : Fuzzy Control. Addisson Wesley, 1997.
- [3] Siler, W. and Ying, H. : Fuzzy Control Theory: The Linear Case. Fuzzy Sets and Systems, 33: 275–290, 1989.
- [4] Wolkenhauer, O. : Data Engineering. http://www.csc.umist.ac.uk/people/wolkenhauer.htm.
- [5] Ying, H. and Siler, W. and Buckley, J.J.: Fuzzy Control Theory: A Nonlinear Case. Automatica, 26 (3): 513–520, 1990. 15

